## NP-Preprocessing

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## Overview

## Introduction

## Reducing CNF Formulae <br> P-Preprocessings <br> NP-Preprocessings

## Combining Preprocessings

## Implicit GDR thanks to Definability

## Knowledge Compilation vs. Preprocessing

Two preprocessing approaches for circumventing the complexity of computationally hard tasks

- knowledge compilation
$\begin{aligned} \text { input: } \Sigma\left(\text { in } \mathcal{L}_{1}\right) \longrightarrow \text { compilation } \rightarrow & \Psi\left(\text { in } \mathcal{L}_{2}\right) \xrightarrow{\text { resolution } \rightarrow \text { output: result }} \\ & \text { input: } \alpha\left(\text { in } \mathcal{L}_{1}\right)\end{aligned}$


## Knowledge Compilation vs. Preprocessing

Two preprocessing approaches for circumventing the complexity of computationally hard tasks

- knowledge compilation

- preprocessing
input: $\Sigma, \alpha\left(\right.$ in $\left.\mathcal{L}_{1}\right)$ preprocessing $\rightarrow \psi\left(\right.$ in $\left.\mathcal{L}_{1}\right) \xrightarrow{\text { resolution } \rightarrow \text { output: result }}$


## Knowledge Compilation vs. Preprocessing

Two preprocessing approaches for circumventing the complexity of computationally hard tasks

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- The two approaches can be combined


## Knowledge Compilation vs. Preprocessing

- Main resemblances:
- making the resolution of the instance computationally easier once the preprocessing step has been achieved
- no guarantee of success


## Knowledge Compilation vs. Preprocessing

- Main resemblances:
- making the resolution of the instance computationally easier once the preprocessing step has been achieved
- no guarantee of success
- Main differences:
- "hard" part vs. "easy" part of the solving process
- handling of the variable part $\alpha$ of the input (does not exist in general, can be preprocessed as well or not)


## Computationally Hard Tasks = ?

- SAT
- Input: a CNF formula $\Sigma$
- Output: 1 if $\Sigma$ is satisfiable, 0 otherwise
- The canonical NP-complete problem!
- \#SAT
- Input: a CNF formula $\Sigma$ (plus eventually a satisfiable term $\alpha$ )
- Output: the number of models of $\Sigma$ (conditioned by $\alpha$ )
- The canonical \#P-complete problem!


## Satisfiability

- $\Sigma \mapsto 1$ if $\Sigma$ has a model, 0 otherwise


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- $\Sigma \mapsto 1$ if $\Sigma$ has a model, 0 otherwise
- $\Sigma=(x \vee y) \wedge(\neg y \vee z)$
- $\Sigma$ is satisfiable since (for instance) 011 is a model of $\Sigma$

Model Counting

$$
\triangleright \Sigma \mapsto\|\Sigma\|=?
$$

## Model Counting

- $\Sigma \mapsto\|\Sigma\|=$ ?
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## Model Counting

- $\Sigma \mapsto\|\Sigma\|=$ ?
- $\Sigma=(x \vee y) \wedge(\neg y \vee z)$
- The models of $\Sigma$ over $\{x, y, z\}$ are :

$$
\begin{aligned}
& 011 \\
& 100 \\
& 101 \\
& 111
\end{aligned}
$$

Model Counting

- $\Sigma \mapsto\|\Sigma\|=$ ?
- $\Sigma=(x \vee y) \wedge(\neg y \vee z)$
- The models of $\Sigma$ over $\{x, y, z\}$ are :

$$
\begin{array}{r}
011 \\
100 \\
101 \\
111 \\
-\|\Sigma\|=4
\end{array}
$$

## Model Counting

- Counting the models of a propositional formula is a key task for a number of problems (especially in AI):
- probabilistic inference
- stochastic planning
- ...


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- Even for subsets of formulae for which SAT is easy (e.g., monotone Krom formulae)!


## Model Counting

- Counting the models of a propositional formula is a key task for a number of problems (especially in AI):
- probabilistic inference
- stochastic planning
- ...
- However \#sat is a computationally hard task: \#P-complete
- Even for subsets of formulae for which SAT is easy (e.g., monotone Krom formulae)!
- The "power" of counting and its complexity are reflected by Toda's theorem:
Seinosuke Toda (Gödel Prize 1998):

$$
\mathrm{PH} \subseteq \mathrm{P}^{\# \mathrm{P}}
$$

## Model Counting

- Many model counters have been developed:
- Exact model counters:
- search-based: Cachet, SharpSAT, etc.,
- compilation-based: C2D, Dsharp, D4, etc.
- ...
- Approximate model counters (SampleCount, etc.)
- ...


## Knowledge Compilation vs. Preprocessing for Model Counting

- knowledge compilation



## Knowledge Compilation vs. Preprocessing for Model Counting

- knowledge compilation
input: $\Sigma$ (in CNF) $\xrightarrow[\text { compilation }]{\longrightarrow} \psi$ (in d-DNNF) $\rightarrow\|\Sigma \mid \alpha\|$
input: $\alpha$ (a consistent term)
- preprocessing
input: $\Sigma, \alpha($ in CNF $) \longrightarrow \boldsymbol{p} \longrightarrow \boldsymbol{\text { m }}$ (in CNF $) \longrightarrow\|\Sigma \mid \alpha\|$


## P-Preprocessings

- Polynomial-time preprocessings
- Can prove useful for SAT and \#sat
- Related to the notion of kernelization: given a parameterized problem $\left(L \subseteq V^{*}, \kappa: V^{*} \rightarrow \mathbb{N}\right)$, a kernelization for $L$ is a polynomial-time algorithm $p$ that takes an instance $\Sigma \subseteq V^{*}$ with parameter $\kappa(\Sigma)$, and maps it to an instance $p(\Sigma) \subseteq V^{*}$ such that $\Sigma \in L$ if and only if $p(\Sigma) \in L$ and the size of $p(\Sigma)$ is upper bounded by $f(\kappa(\Sigma))$ for some computable function $f$
- If $L$ is decidable, then $L$ is fixed-parameter tractable for parameter $\kappa($.$) if and only if L$ has a kernelization


## Dozens of P-Preprocessings

- Vivification (VI) and a light form of it, called Occurrence Elimination (OE),
- Gate Detection and Replacement (GDR)
- Pure Literal Elimination (PLE)
- Variable Elimination (VE)
- Blocked Clause Elimination (BCE)
- Covered Clause Elimination (CCE)
- Failed Literal Elimination (FLE)
- Self-Subsuming Resolution (SSR)
- Hidden Literal Elimination (HLE)
- Subsumption Elimination (SE)
- Hidden Subsumption Elimination (HSE)
- Asymmetric Subsumption Elimination (ASE)
- Tautology Elimination (TE)
- Hidden Tautology Elimination (HTE)
- Asymmetric Tautology Elimination (ATE)


## Use in State-of-the-Art SAT Solvers

- Glucose (exploits the SatELite preprocessor)
- Lingeling (has an internal preprocessor)
- Riss (use of the Coprocessor preprocessor)
- ...


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## Reducing CNF Formulae

## P-Preprocessings

NP-Preprocessings

Combining Preprocessings

## Implicit GDR thanks to Definability

## The CNF Input Assumption

SAT solvers and model counters typically considers CNF inputs

- The CNF assumption is not restrictive
- Every circuit $\Psi$ can be turned into a CNF formula $\Sigma$ in linear time
- The translation requires the introduction of new variables $Y=\left\{y_{1}, \ldots,\right\}$ and preserves
- the queries over the alphabet of the input circuit (Plaisted/Greenbaum)

$$
\Psi \equiv \exists Y . \Sigma
$$

- and the number of models of the input (Tseitin)


## Translation into CNF: Tseitin

$$
\begin{aligned}
& \Psi=\left(x_{1} \wedge \bar{x}_{2}\right) \vee\left(x_{2} \wedge x_{3}\right) \\
& T(\Psi) \equiv\left(y_{1} \vee y_{2}\right) \wedge\left(y_{1} \Leftrightarrow\left(x_{1} \wedge \bar{x}_{2}\right)\right) \wedge\left(y_{2} \Leftrightarrow\left(x_{2} \wedge x_{3}\right)\right) \\
& T(\Psi)= \\
& y_{1} \vee y_{2} \\
& \bar{y}_{1} \vee x_{1} \\
& \bar{y}_{1} \vee \bar{x}_{2} \\
& y_{1} \vee \bar{x}_{1} \vee x_{2} \\
& \bar{y}_{2} \vee x_{2} \\
& \bar{y}_{2} \vee x_{3} \\
& y_{2} \vee \bar{x}_{2} \vee \bar{x}_{3}
\end{aligned}
$$

## Translation into CNF: Plaisted/Greenbaum

$$
\begin{aligned}
& \Psi=\left(x_{1} \wedge \bar{x}_{2}\right) \vee\left(x_{2} \wedge x_{3}\right) \\
& P G(\Psi) \equiv\left(y_{1} \vee y_{2}\right) \wedge\left(y_{1} \Rightarrow\left(x_{1} \wedge \bar{x}_{2}\right)\right) \wedge\left(y_{2} \Rightarrow\left(x_{2} \wedge x_{3}\right)\right) \\
& P G(\Psi)= \\
& \quad y_{1} \vee y_{2} \\
& \bar{y}_{1} \vee x_{1} \\
& \bar{y}_{1} \vee \bar{x}_{2} \\
& \bar{y}_{2} \vee x_{2} \\
& \bar{y}_{2} \vee x_{3}
\end{aligned}
$$

## Reducing What?

CNF $\Sigma \mapsto \operatorname{CNF} p(\Sigma)$

- What are the connections between $\Sigma$ and $p(\Sigma)$ ?
- Removing clauses from $\Sigma$
- Removing literals in the clauses of $\Sigma$
- Updating $\Sigma$ in another way


## Looking for IES or Minimal CNF is often too Expensive

- A clause $\delta$ of a CNF $\Sigma$ is redundant if and only if $\Sigma \backslash\{\delta\} \models \delta$
- A CNF $\Sigma$ is irredundant if and only if it does not contain any redundant clause
- A subset $\Sigma^{\prime}$ of a CNF $\Sigma$ is an irredundant equivalent subset (IES) of $\Sigma$ if and only if $\Sigma^{\prime}$ is irredundant and $\Sigma^{\prime} \equiv \Sigma$
- Deciding whether a CNF $\Sigma$ is irredundant is NP-complete
- Deciding whether a CNF $\Sigma^{\prime}$ is an irredundant equivalent subset (IES) of a CNF $\Sigma$ is $\mathrm{D}^{p}$-complete
- Given an integer $k$, deciding whether a CNF $\Sigma$ has an IES of size at most $k$ is $\sum_{2}^{p}$-complete
- Given an integer $k$, deciding whether there exists a CNF formula $\Sigma^{\prime}$ with at most $k$ literals (or with at most $k$ clauses) equivalent to a given $\operatorname{CNF} \Sigma$ is $\Sigma_{2}^{p}$-complete


## Redundancy can be Useful!

Redundancy can be helpful (favoring the unit propagation power by adding empowering clauses)

- $x_{3} \vee x_{4}$ is a logical consequence of

$$
\begin{aligned}
\Sigma= & x_{1} \vee x_{2} \vee x_{3} \\
& \bar{x}_{1} \vee x_{2} \vee x_{4} \\
& x_{1} \vee \bar{x}_{2} \vee x_{3} \\
& \bar{x}_{1} \vee \bar{x}_{2} \vee x_{4}
\end{aligned}
$$

- Assuming $\bar{x}_{3} \wedge \bar{x}_{4}$, there is no unit refutation from $\Sigma$
- If $x_{3} \vee x_{4}$ (or $x_{2} \vee x_{3} \vee x_{4}$ or $\bar{x}_{2} \vee x_{3} \vee x_{4}$ ) is added to $\Sigma$, a unit refutation from $\Sigma$ under the assumption $\bar{x}_{3} \wedge \bar{x}_{4}$ exists
- Learning clauses is a key ingredient for efficient SAT solvers based on the CDCL architecture!


## Preserving What?

- Logical equivalence
- Number of models
- Satisfiability
- ...


## Levels of Preservation

$A \rightarrow B$ : if a preprocessing $p$ preserves $A$, then $p$ preserves $B$

Logical equivalence


## Properties of Preprocessings

- Level of preservation (logical equivalence, number of models, satisfiability)
- Confluence: is $p(\Sigma)$ sensitive w.r.t. the way clauses and literals in them are listed in $\Sigma$ ?
- Projectiveness: do we have $p(p(\Sigma))=p(\Sigma)$ ? (important to decide whether iterating $p$ makes sense or not)


## Measuring the Impact of a Preprocessing

Several measures for the reduction achieved can be considered:

- The number of variables in the input CNF $\Sigma$
- The size of $\Sigma$ (the number of literals or the number of clauses in it)
- The treewidth of the primal graph of $\Sigma$
- The value of other structural parameters for $\Sigma$
- ...


## Example: Subsumption Elimination

A clause $\delta_{1}$ subsumes a clause $\delta_{2}$
if every literal of $\delta_{1}$ is a literal of $\delta_{2}$
$S E:\left(x_{1} \vee x_{2}\right) \wedge\left(x_{1} \vee x_{2} \vee \bar{x}_{3}\right) \mapsto x_{1} \vee x_{2}$

- Preserves logical equivalence
- Hence preserves the number of models of the input (over the original alphabet), and its satisfiability
- Is confluent and projective
- \#var $(\Sigma) \geq \# \operatorname{var}(\operatorname{SE}(\Sigma))$
- \#lit $(\Sigma) \geq \# \operatorname{lit}(\operatorname{SE}(\Sigma))$
- $t w(\Sigma) \geq t w(\operatorname{SE}(\Sigma))$


## Example: Pure Literal Elimination

A literal / is pure in $\Sigma$ if every occurrence of the corresponding variable has the same polarity

$$
P L E:\left(x_{1} \vee x_{2} \vee x_{3}\right) \wedge\left(\bar{x}_{2} \vee \bar{x}_{3}\right) \mapsto \bar{x}_{2} \vee \bar{x}_{3}
$$

- Preserves the satisfiability of the input
- Does not preserve its number of models or logical equivalence
- Is confluent and projective
- \#var $(\Sigma) \geq \# \operatorname{var}(\operatorname{PLE}(\Sigma))$
- \#lit $(\Sigma) \geq \# \operatorname{lit}(\operatorname{PLE}(\Sigma))$
- $t w(\Sigma) \geq t w(\operatorname{PLE}(\Sigma))$


## Estimating the Amount of Reduction Achieved

- For each $p$, the impact of $p$ is evaluated empirically (and quantitatively) on 182 CNF instances from the SAT LIBrary, gathered into 9 data sets, as follows:
- Bayesian networks (60)
- BMC (11) (Bounded Model Checking)
- Circuit (28)
- Configuration (12)
- Handmade (28)
- Planning (17)
- Random (13)
- Scheduling (6)
- Qif (7) (Quantitative Information Flow analysis - security)
- Cluster of Intel Xeon E5-2643 (3.30 GHz) processors with 32 GiB RAM on Linux CentOS
- Time-out $=1 \mathrm{~h}$
- Memory-out $=7.6 \mathrm{GiB}$


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## Occurrence Reduction (OR)

- Occurrence reduction aims to remove some literals in the clauses of $\Sigma$
- In order to determine whether a literal $\ell_{j+1}$ can be removed from a clause $\alpha$ of $\Sigma$, the approach consists in determining whether the clause which coincides with $\alpha$ except that $\ell_{j+1}$ has been replaced by $\sim \ell_{j+1}$ is a logical consequence of $\Sigma$
- This is done by determining whether $\Sigma \wedge \ell_{j+1} \wedge \sim \ell_{1} \wedge \ldots \wedge \sim \ell_{j}$ is contradictory
- When this is the case, $\ell_{j+1}$ can be removed from $\alpha$ without questioning logical equivalence
- bcp is used as an incomplete yet efficient method to solve the entailment problem


## Occurrence Reduction (OR)

Algorithm 1: OR Occurrence reduction
input : a CNF formula $\Sigma$
output: a CNF formula equivalent to $\Sigma$
$1 \mathcal{L} \leftarrow \operatorname{sort}(\operatorname{Lit}(\Sigma))$;
2 foreach $\ell \in \mathcal{L}$ do
3 foreach $\alpha \in \Sigma$ s.t. $\ell \in \alpha$ do
4 if $\emptyset \in \operatorname{bcp}(\Sigma \cup\{\ell\} \cup\{\sim(\alpha \backslash\{\ell\})\})$ then $\Sigma \leftarrow(\Sigma \backslash\{\alpha\}) \cup\{\alpha \backslash\{\ell\}\} ;$

5 return $\Sigma$

## Occurrence Reduction (OR): Example

$$
\begin{aligned}
& \Sigma= \\
& a \vee f \quad a \vee b \vee c \\
& b \vee d \vee e \quad c \vee \neg d \vee e \\
& b \vee d \vee \neg e \quad c \vee \neg d \vee \neg e \\
& \mathcal{L}=(b, c, e, \neg e, d, \neg d, a, f, \neg a, \neg b, \neg c) \\
& \mathrm{OR}(\Sigma)= \\
& a \vee f \quad b \vee c \\
& b \vee d \quad c \vee \neg d \\
& b \vee d \quad c \vee \neg d
\end{aligned}
$$

$a \vee b \vee c$ is reduced to $b \vee c$ and every occurrence of $e$ has been removed

## Properties of OR

- Preserves logical equivalence
- Neither is confluent nor is projective
- \#var $(\Sigma) \geq \# \operatorname{var}(\mathrm{OR}(\Sigma))$
- \#lit $(\Sigma) \geq \# \operatorname{lit}(\operatorname{OR}(\Sigma))$
- $t w(\Sigma) \geq t w(\mathrm{OR}(\Sigma))$


## OR: Reduction of the Number of Variables



Figure: Comparing \#var $(\Sigma)$ with $\# \operatorname{var}(\mathrm{OR}(\Sigma))$.

## OR: Reduction of the Size



Figure: Comparing \#lit( $\Sigma$ ) with \# $\operatorname{lit}(\operatorname{OR}(\Sigma))$.

## OR: Reduction of the Treewidth



Figure: Comparing $t w(\Sigma)$ with $t w(\operatorname{OR}(\Sigma))$.

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## Vivification (VI)

- Vivification aims to reduce $\Sigma$, i.e., to remove some clauses and some literals in $\Sigma$ while preserving equivalence
- Given a clause $\alpha=\ell_{1} \vee \ldots \vee \ell_{k}$ of $\Sigma$ two rules are used in order to determine whether $\alpha$ can be removed from $\Sigma$ or simply shortened
- Let $\alpha^{\prime}$ be any subclause of $\alpha$
- On the one hand, if for any $j \in 0, \ldots, k-1$, one can prove using bcp that $\Sigma \backslash\{\alpha\} \vDash \alpha^{\prime}$, then for sure $\alpha$ is entailed by $\Sigma \backslash\{\alpha\}$ so that $\alpha$ can be removed from $\Sigma$
- On the other hand, if one can prove using bcp that $\Sigma \backslash\{\alpha\} \models \alpha^{\prime} \vee \sim \ell_{j+1}$, then $\ell_{j+1}$ can be removed from $\alpha$ without questioning equivalence


## Vivification (VI)

```
Algorithm 2: VI input : a CNF formula \(\Sigma\) output: a CNF formula equivalent to \(\Sigma\)
```

1 foreach $\alpha \in \Sigma$ do

```
\(2 \quad \Sigma \leftarrow \Sigma \backslash\{\alpha\}\);
\(3 \quad \alpha^{\prime} \leftarrow \perp\);
\(4 \quad \mathcal{I} \leftarrow \operatorname{bcp}(\Sigma)\);
\(5 \quad\) while \(\exists \ell \in \alpha\) s.t. \(\sim \ell \notin \mathcal{I}\) and \(\alpha^{\prime} \neq \top\) do
\(6 \quad \alpha^{\prime} \leftarrow \alpha^{\prime} \vee \ell\);
\(7 \quad \mathcal{I} \leftarrow \operatorname{bcp}\left(\Sigma \cup\left\{\sim \alpha^{\prime}\right\}\right)\);
\(8 \quad\) if \(\emptyset \in \mathcal{I}\) then \(\alpha^{\prime} \leftarrow \mathrm{T}\);
\(9 \quad \Sigma \leftarrow \Sigma \cup\left\{\alpha^{\prime}\right\}\);
10 return \(\Sigma\)
```


## Vivification (VI): Example

$$
\begin{aligned}
\Sigma= & \\
& a \vee b \vee c \vee d \\
& a \vee b \vee c \\
& a \vee \neg d
\end{aligned}
$$

Assume that the variables are processed w.r.t. the ordering
$d<c<b<a$
$\mathrm{VI}(\Sigma)=$
$a \vee b \vee c$
$a \vee \neg d$
The effect of VI on $\Sigma$ is to eliminate the first clause $a \vee b \vee c \vee d$

## Properties of VI

- Preserves logical equivalence
- Neither is confluent nor is projective
- \#var $(\Sigma) \geq \# \operatorname{var}(\operatorname{VI}(\Sigma))$
- \#lit $(\Sigma) \geq \# \operatorname{lit}(\operatorname{VI}(\Sigma))$
- $t w(\Sigma) \geq t w(\mathrm{VI}(\Sigma))$


## VI: Reduction of the Number of Variables



Figure: Comparing \#var( $\Sigma$ ) with \#var(VI( $\Sigma)$ ).

## VI: Reduction of the Size



Figure: Comparing $\# \operatorname{lit}(\Sigma)$ with $\# \operatorname{lit}(\operatorname{VI}(\Sigma))$.

## VI: Reduction of the Treewidth



Figure: Comparing $t w(\Sigma)$ with $t w(\operatorname{VI}(\Sigma))$.

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Literal Equivalence (LE)AND/OR Gate Equivalence (AG)XOR Gate Equivalence (XG)
NP-Preprocessings
Combining Preprocessings

## Gate Detection and Replacement

A gate of $\Sigma$ is a circuit $\ell \Leftrightarrow \beta$ such that $\Sigma \models \ell \Leftrightarrow \beta$
$\Sigma$ and $\Sigma[\ell \leftarrow \beta]$ have the same number of models (but are not logically equivalent in general)

$$
\Sigma=\quad \begin{aligned}
& \bar{x} \vee u \vee v \\
& \bar{x} \vee \bar{y} \vee u \\
& \bar{x} \vee \bar{z} \vee u \\
& x \vee \bar{u} \\
& y \vee z \vee \bar{u}
\end{aligned}
$$

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& \\
& \bar{x} \vee \bar{y} \vee u \\
& \bar{x} \vee \bar{z} \vee u \\
& \\
& x \vee \bar{u} \\
& \\
& y \vee z \vee \bar{u}
\end{aligned} \quad u \leftrightarrow(x \wedge(y \vee z))
$$

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$\Sigma$ and $\Sigma[\ell \leftarrow \beta]$ have the same number of models (but are not logically equivalent in general)

$$
\begin{array}{ll} 
& \bar{x} \vee u \vee v \\
& \bar{x} \vee \bar{y} \vee u \\
& \bar{x} \vee \bar{z} \vee u \\
& x \vee \bar{u} \\
& y \vee z \vee \bar{u} \\
\Sigma \equiv & \\
(\bar{x} \vee u \vee v) \wedge(u \leftrightarrow(x \wedge(y \vee z))) & \\
& \\
& \text { detection }
\end{array}
$$

## Gate Detection and Replacement

A gate of $\Sigma$ is a circuit $\ell \Leftrightarrow \beta$ such that $\Sigma \models \ell \Leftrightarrow \beta$
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$$
\begin{aligned}
& \bar{x} \vee u \vee v \\
& \bar{x} \vee \bar{y} \vee u \\
& \Sigma=\bar{x} \vee \bar{z} \vee u \quad u \leftrightarrow(x \wedge(y \vee z)) \\
& x \vee \bar{u} \\
& y \vee z \vee \bar{u} \\
& \text { detection } \\
& \text { replacement }
\end{aligned}
$$

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\begin{aligned}
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& x \vee \bar{u} \\
& y \vee z \vee \bar{u} \\
& \text { detection } \\
& \text { replacement } \\
& \text { normalization }
\end{aligned}
$$

## Gate Detection and Replacement

A gate of $\Sigma$ is a circuit $\ell \Leftrightarrow \beta$ such that $\Sigma \models \ell \Leftrightarrow \beta$
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$$
\Sigma=\begin{aligned}
& \bar{x} \vee u \vee v \\
& \bar{x} \vee \bar{y} \vee u \\
& \bar{x} \vee \bar{z} \vee u \\
& \\
& x \vee \bar{u} \\
& \\
& y \vee z \vee \bar{u}
\end{aligned} \quad u \leftrightarrow(x \wedge(y \vee z))
$$

detection replacement normalization

$$
\|\Sigma\|=\|\Sigma[u \leftarrow(x \wedge(y \vee z))]\|=\|\bar{x} \vee y \vee z \vee v\|=15
$$

## Gate Detection and Replacement

- Gate detection and replacement proves to be a valuable preprocessing
- Only specific gates are sought for (literal equivalence, AND/OR gates, XOR gates)
- The replacement $\Sigma[\ell \leftarrow \beta]$ requires to turn the resulting formula into CNF
- It is implemented only if it it does not lead to increase the size of the input (a "small" increase can also be accepted)
- bcp (instead of a "full" sat solver) is used for efficiency reasons


## Literal Equivalence (LE)

- Literal equivalence aims to detect equivalences between literals using bcp
- For each literal $\ell$, all the literals $\ell^{\prime}$ which can be found equivalent to $\ell$ using bcp are replaced by $\ell$ in $\Sigma$
- Taking advantage of bcp makes it more efficient than a "syntactic detection" (if two binary clauses stating an equivalence between two literals $\ell$ and $\ell^{\prime}$ occur in $\Sigma$, then those literals are found equivalent using bcp, but the converse does not hold)


## Literal Equivalence (LE)

```
Algorithm 3: LE
input : a CNF formula \(\Sigma\)
output: a CNF formula \(\Phi\) such that \(\|\Phi\|=\|\Sigma\|\)
\(1 \Phi \leftarrow \Sigma\); Unmark all variables of \(\Phi\);
2 while \(\exists \ell \in \operatorname{Lit}(\Phi)\) s.t. \(\operatorname{var}(\ell)\) is not marked do
    // detection
        mark var \((\ell)\);
        \(\mathcal{P}_{\ell} \leftarrow \mathrm{bcp}(\Phi \cup\{\ell\}) ;\)
        \(\mathcal{N}_{\ell} \leftarrow \mathrm{bcp}(\Phi \cup\{\sim \ell\}) ;\)
        \(\Gamma \leftarrow\left\{\ell \leftrightarrow \ell^{\prime} \mid \ell^{\prime} \neq \ell\right.\) and \(\ell^{\prime} \in \mathcal{P}_{\ell}\) and \(\left.\sim \ell^{\prime} \in \mathcal{N}_{\ell}\right\}\);
        // replacement
        foreach \(\ell \leftrightarrow \ell^{\prime} \in \Gamma\) do
            replace \(\ell\) by \(\ell^{\prime}\) in \(\Phi\);
9 return \(\Phi\)
```


## Literal Equivalence (LE): Example

$$
\begin{array}{rlrl}
\Sigma= & & \\
& a \vee b \vee c \vee \neg d & & \neg a \vee \neg b \vee \neg c \vee d \\
& a \vee b \vee \neg c & & \neg a \vee \neg b \vee c \\
& \neg a \vee b & & a \vee \neg b \\
& \neg e \vee \neg f \vee h & & e \vee f \vee g \\
& e \vee \neg g & & \neg e \vee \neg h
\end{array}
$$

Assume that the variables of $\Sigma$ are considered in the following ordering: $a<b<c<d<e<f<g<h$

The equivalences $(a \Leftrightarrow b) \wedge(b \Leftrightarrow c) \wedge(c \Leftrightarrow d) \wedge(e \Leftrightarrow \neg f)$ are detected

$$
\begin{aligned}
& \mathrm{LE}(\Sigma)= \\
& \quad \neg f \vee \neg g \quad f \vee \neg h
\end{aligned}
$$

## Properties of LE

- Preserves the number of models (but not logical equivalence)
- Neither is confluent nor is projective
- \#var $(\Sigma) \geq \# \operatorname{var}(\mathrm{LE}(\Sigma))$
- \#lit $(\Sigma) \geq \# \operatorname{lit}(\operatorname{LE}(\Sigma))$
- $t w(\Sigma) \nsupseteq t w(\operatorname{LE}(\Sigma))$


## LE: Reduction of the Number of Variables



Figure: Comparing \#var( $\Sigma$ ) with \#var $(\operatorname{LE}(\Sigma))$.

## LE: Reduction of the Size



Figure: Comparing $\# \operatorname{lit}(\Sigma)$ with $\# \operatorname{lit}(\operatorname{LE}(\Sigma))$.

## LE: Reduction of the Treewidth



## Figure: Comparing $t w(\Sigma)$ with $t w(\operatorname{LE}(\Sigma))$.

## AND/OR Gate Equivalence (AG)

- AND/OR gate equivalence aims to detect equivalences $\ell_{i} \Leftrightarrow \beta_{i}$ where $\beta_{i}$ is a conjunction or a disjunction of literals
- bcp is used in this objective
- Unlike previous approaches based on pattern matching (i.e., when one looks for clauses encoding an AND gate or an OR gate), using bcp makes it more efficient (if clauses stating the presence of an AND/OR gate occur in $\Sigma$, then AG will find the gate - or a "subsuming" one - but the converse is not true)


## AND/OR Gate Equivalence (AG)

```
Algorithm 4: AG
input : a CNF formula \(\Sigma\)
output : a CNF formula \(\Phi\) such that \(\|\Phi\|=\|\Sigma\|\)
\(\Phi \leftarrow \Sigma\);
// detection
\(\Gamma \leftarrow \emptyset\); Unmark all literals of \(\Phi\);
while \(\exists \ell \in \operatorname{Lit}(\Phi)\) s.t. \(\ell\) is not marked do
        mark \(\ell\);
        \(\mathcal{P}_{\ell} \leftarrow(\operatorname{bcp}(\Phi \cup\{\ell\}) \backslash(\operatorname{bcp}(\Phi) \cup\{\ell\})) \cup\{\sim \ell\} ;\)
        if \(\emptyset \in \operatorname{bcp}\left(\Phi \cup \mathcal{P}_{\ell}\right)\) then
                        let \(\mathcal{C}_{\ell} \subseteq \mathcal{P}_{\ell}\) s.t. \(\emptyset \in \operatorname{bcp}\left(\Phi \cup \mathcal{C}_{\ell}\right)\) and \(\sim \ell \in \mathcal{C}_{\ell}\);
                        \(\Gamma \leftarrow \Gamma \cup\left\{\ell \leftrightarrow \bigwedge_{\ell^{\prime} \in \mathcal{C}_{\ell} \backslash\left\{\sim_{\ell}\right\}} \ell^{\prime}\right\} ;\)
// replacement
while \(\exists \ell \leftrightarrow \beta \in \Gamma\) st. \(|\beta|<\max A\) and \(|\Phi[\ell \leftarrow \beta]| \leq|\Phi|\) do
    \(\Phi \leftarrow \Phi[\ell \leftarrow \beta] ;\)
    \(\Gamma \leftarrow \Gamma[\ell \leftarrow \beta] ;\)
return \(\Phi\)
```


## AND/OR Gate Equivalence (AG): Example

$$
\begin{aligned}
\Sigma= & & & \\
& a \vee b \vee c \vee d & & \neg a \vee \neg b \\
& \neg a \vee b \vee \neg c & & \neg a \vee b \vee c \vee \neg d \\
& \neg a \vee e & & a \vee f
\end{aligned}
$$

Assume that the literals are processed w.r.t. the literal ordering: $a<\neg a<b<\neg b<c<\neg c<d<\neg d<e<\neg e<f<\neg f$

The gate $a \Leftrightarrow(\neg b \wedge \neg c \wedge \neg d)$ is detected and $a$ is replaced by its definition

$$
\begin{array}{ll}
\mathrm{AG}(\Sigma)= & \\
\quad b \vee c \vee d \vee e & \neg b \vee f \\
\neg c \vee f & \neg d \vee f
\end{array}
$$

## Properties of AG

- Preserves the number of models (but not logical equivalence)
- Neither is confluent nor is projective
- $\# \operatorname{var}(\Sigma) \geq \# \operatorname{var}(\mathrm{AG}(\Sigma))$
- \#lit $(\Sigma) \geq \# \operatorname{lit}(\operatorname{AG}(\Sigma))$
- $t w(\Sigma) \nsupseteq t w(\operatorname{AG}(\Sigma))$


## AG: Reduction of the Number of Variables



Figure: Comparing \#var( $\Sigma$ ) with \#var(AG( $\Sigma)$ ).

## AG: Reduction of the Size



Figure: Comparing $\# \operatorname{lit}(\Sigma)$ with $\# \operatorname{lit}(\operatorname{AG}(\Sigma))$.

## AG: Reduction of the Treewidth



Figure: Comparing $t w(\Sigma)$ with $\operatorname{tw}(\operatorname{AG}(\Sigma))$.

## XOR Gate Equivalence (XG)

- XOR gate detection and replacement aims to detect equivalences $\ell_{i} \Leftrightarrow \chi_{i}$ where $\chi_{i}$ is a XOR disjunction of literals
- The detection is "syntactic" (XOR disjunctions of literals containing a limited number of literals are targeted)
- The resulting set of gates, which can be viewed as a set of XOR clauses since $\ell_{i} \leftrightarrow \chi_{i}$ is equivalent to $\sim \ell_{i} \oplus \chi_{i}$, is turned into reduced row echelon form using Gauss algorithm
- Every $\ell_{i}$ is replaced by its definition $\chi_{i}$ in $\Sigma$, provided that the normalization it involves does not generate "large" clauses


## XOR Gate Equivalence (XG)

Algorithm 5: XG
input : a CNF formula $\Sigma$
output: a CNF formula $\Phi$ such that $\|\Phi\|=\|\Sigma\|$
$1 \Phi \leftarrow \Sigma$;
// detection
$2 \Gamma \leftarrow\{$ XOR clauses syntactically detected $\}$;
// Gaussian elimination
$3 \Gamma \leftarrow \operatorname{Gauss}\left(\left\{\ell_{1} \leftrightarrow \chi_{1}, \ell_{2} \leftrightarrow \chi_{2}, \ldots, \ell_{k} \leftrightarrow \chi_{k}\right\}\right)$
// replacement
4 for $i \leftarrow 1$ to $k$ do
$5\left\lfloor\right.$ if $\nexists \alpha \in \Phi\left[\ell_{i} \leftarrow \chi_{i}\right]$ s.t. $|\alpha|>\max X$ then $\Phi \leftarrow \Phi\left[\ell_{i} \leftarrow \chi_{i}\right]$;
6 return $\Phi$

## XOR Gate Equivalence (XG): Example

$$
\begin{array}{rlrl}
\Sigma= & & b \vee d & \\
& \checkmark \vee \neg d \\
& \neg a \vee \neg b \vee c & & a \vee \neg b \vee \neg c \\
& \neg a \vee b \vee \neg c & & a \vee b \vee c \\
& b \vee e & & a \vee f
\end{array}
$$

Suppose that the two XOR gates $b \oplus d$ and $a \oplus b \oplus c$ are detected successively

$$
\begin{aligned}
& \mathrm{XG}(\Sigma)= \\
& \quad \neg d \vee e \\
& \quad \neg c \vee \neg d \vee f
\end{aligned} \quad c \vee d \vee f
$$

The first six clauses which participate in the XOR gates are made valid through the replacement, the clause $b \vee e$ is replaced by $\neg d \vee e$, and the clause $a \vee f$ is replaced by the two clauses $c \vee d \vee f$ and $\neg c \vee \neg d \vee f$

## Properties of XG

- Preserves the number of models (but not logical equivalence)
- Neither is confluent nor is projective
- \#var $(\Sigma) \geq \# \operatorname{var}(\mathrm{XG}(\Sigma))$
- \#lit $(\Sigma) \geq \# \operatorname{lit}(\mathrm{XG}(\Sigma))$
- $t w(\Sigma) \nsupseteq t w(\mathrm{XG}(\Sigma))$


## XG: Reduction of the Number of Variables



Figure: Comparing \#var( $\Sigma$ ) with \#var $(\mathrm{XG}(\Sigma))$.

## XG: Reduction of the Size



Figure: Comparing \# lit( $\Sigma$ ) with $\# \operatorname{lit}(\mathrm{XG}(\Sigma))$.

## XG: Reduction of the Treewidth



Figure: Comparing $t w(\Sigma)$ with $t w(\mathrm{XG}(\Sigma))$.

## Overview

## Introduction

## Reducing CNF Formulae

P-Preprocessings

NP-Preprocessings
Backbone Identification (BI)

## Combining Preprocessings

Implicit GDR thanks to Definability

## NP-Preprocessings

- Taking advantage of SAT solvers for simplifying the input CNF during a preprocessing step
- An option that makes sense when tackling problems which are seemingly located "above NP" (like \#sAT)
- Backbone identification
- Definability


## Overview

## Introduction

## Reducing CNF Formulae

P-Preprocessings

NP-Preprocessings
Backbone Identification (BI)

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Implicit GDR thanks to Definability

## Backbone Identification (BI)

- The backbone of a CNF formula $\Sigma$ is the set of all literals which are implied by $\Sigma$ when $\Sigma$ is satisfiable, and is the empty set otherwise
- The purpose of the $B I$ preprocessing is to make the backbone $B$ of the input CNF formula $\Sigma$ explicit, to conjoin it to $\Sigma$, and to use bcp (Boolean Constraint Propagation) on the resulting set of clauses


## Backbone Identification (BI)

Algorithm 6: BI Backbone Identification
input : a CNF formula $\Sigma$
output: the CNF $\operatorname{bcp}(\Sigma \cup B)$, where $\mathcal{B}$ is the backbone of $\Sigma$
$1 \mathcal{B} \leftarrow \emptyset$;
$2 \mathcal{I} \leftarrow \operatorname{solve}(\Sigma)$;
3 while $\exists \ell \in \mathcal{I}$ s.t. $\ell \notin \mathcal{B}$ do
$4 \quad \mathcal{I}^{\prime} \leftarrow \operatorname{solve}(\Sigma \cup\{\sim \ell\})$;
$5 \quad$ if $\mathcal{I}^{\prime}=\emptyset$ then $\mathcal{B} \leftarrow \mathcal{B} \cup\{\ell\}$ else $\mathcal{I} \leftarrow \mathcal{I} \cap \mathcal{I}^{\prime}$;
6 return $\operatorname{bcp}(\Sigma \cup \mathcal{B})$

## Backbone Identification (BI): Example

$$
\begin{aligned}
\Sigma= & \\
& a \vee b \\
& \neg a \vee b \\
& \neg b \vee c \\
& c \vee d \\
& \neg c \vee e \vee f \\
& f \vee \neg g
\end{aligned}
$$

The backbone of $\Sigma$ is equal to $B=\{b, c\}$

$$
\begin{gathered}
\mathrm{BI}(\Sigma)= \\
b \\
c \\
e \vee f \\
f \vee \neg g
\end{gathered}
$$

## Properties of BI

- Preserves logical equivalence
- Is confluent and projective
- \#var $(\Sigma) \geq \# \operatorname{var}(\mathrm{BI}(\Sigma))$
- \#lit $(\Sigma) \geq \# \operatorname{lit}(\operatorname{BI}(\Sigma))$
- $t w(\Sigma) \geq t w(\operatorname{BI}(\Sigma))$


## BI: Reduction of the Number of Variables



Figure: Comparing $\# \operatorname{var}(\Sigma)$ with $\# \operatorname{var}(\mathrm{BI}(\Sigma))$.

## BI: Reduction of the Size



Figure: Comparing \# lit $(\Sigma)$ with $\# \operatorname{lit}(\mathrm{BI}(\Sigma))$.

## BI: Reduction of the Treewidth



Figure: Comparing $t w(\Sigma)$ with $t w(\operatorname{BI}(\Sigma))$.

## Overview

## Introduction

## Reducing CNF Formulae

P-Preprocessings

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## Putting the Elementary Preprocessings Together

The pmc preprocessor

- Iteration makes sense
- The combination to be chosen depends on what is expected to be preserved
- eq corresponds to the parameter assignment of pmc where opt $V=o p t B=o p t O=1$ and $o p t G=0$. It is equivalence-preserving.
- \#eq corresponds to the parameter assignment of pmc where opt $V=o p t B=o p t O=1$ and $o p t G=1$. This combination is guaranteed only to preserve the number of models of the input.


## The pmc Preprocessor

```
Algorithm 7: pmc
input : a CNF formula \(\Sigma\)
output: a CNF formula \(\Phi\) such that \(\|\Phi\|=\|\Sigma\|\)
\(\Phi \leftarrow \Sigma\);
if optB then \(\Phi \leftarrow \mathrm{BI}(\Phi)\);
\(i \leftarrow 0\);
4 while \(i<n u m\) Tries do
\(5 \quad i \leftarrow i+1\);
if optO then \(\Phi \leftarrow \operatorname{OR}(\Phi)\);
if opt \(G\) then \(\Phi \leftarrow \mathrm{XG}(\operatorname{AG}(\operatorname{LE}(\Phi)))\);
if opt \(V\) then \(\Phi \leftarrow V I(\Phi)\);
    if fixpoint then break;
10 return \(\Phi\)
```


## eq: Reduction of the Number of Variables



Figure: Comparing \#var( $\Sigma$ ) with \#var $(e q(\Sigma))$.

## eq: Reduction of the Size



Figure: Comparing \#lit( $\Sigma$ ) with \# lit $(e q(\Sigma))$.

## eq: Reduction of the Treewidth



Figure: Comparing $t w(\Sigma)$ with $t w(e q(\Sigma))$.

## \#eq: Reduction of the Number of Variables



Figure: Comparing \#var( $\Sigma$ ) with \#var(\#eq( $\Sigma)$ ).

## \#eq: Reduction of the Size



## Figure: Comparing \#lit( $\Sigma$ ) with \#lit(\#eq( $\Sigma)$ ).

## \#eq: Reduction of the Treewidth



Figure: Comparing $t w(\Sigma)$ with $t w(\# e q(\Sigma))$.

## Improving Model Counting?

- Reducing \#var( $\Sigma$ ), \#lit( $\Sigma$ ), and $t w(\Sigma)$ is a priori valuable for improving the model counting task (computationally speaking)
- But it could be the case that the preprocessings used are too much time-demanding and that they do not really lead to easier instances but concentrate their difficulty instead...
- Some large-scale experiments are needed to determine whether some improvements are actually achieved
- Cachet, SharpSAT, C2D, and Dsharp are used downstream
- C2D and Dsharp are used as compilers (since the objective is to preserve the information given in the input, only eq is considered as an admissible preprocessing combination)


## Empirical Setting

- 1449 CNF instances from the SAT LIBrary
- 9 data sets: BMC (18), Circuit (68), Qif (7), Planning (34), Random (105), Scheduling (6), Handmade (58), Configuration (35), Bayesian networks (1118)
- Cluster of Intel Xeon E5-2643 (3.30 GHz) processors with 32 GiB RAM on Linux CentOS
- Time-out $=1 \mathrm{~h}$
- Memory-out $=7.6 \mathrm{GiB}$


## Efficiency of eq



Number of instances

## Efficiency of \#eq



Number of instances

## Impact of eq on Cachet



Figure: Comparison of the computation times needed to count the number of models of an instance using Cachet, when no preprocessing is used vs. when the eq combination has been applied first.

## Impact of \#eq on Cachet



Figure: Comparison of the computation times needed to count the number of models of an instance using Cachet, when no preprocessing is used vs. when the \#eq combination has been applied first.

## Impact of eq on SharpSAT



Figure: Comparison of the computation times needed to count the number of models of an instance using SharpSAT, when no preprocessing is used vs. when the eq combination has been applied first.

## Impact of \#eq on SharpSAT



Figure: Comparison of the computation times needed to count the number of models of an instance using SharpSAT, when no preprocessing is used vs. when the \#eq combination has been applied first.

## Impact of eq on C2D (compilation times)



Figure: Comparisons of the compilation times of C2D, when no preprocessing is used vs. when the eq combination has been applied first.

## Impact of eq on C2D (sizes of the compiled forms)



Figure: Comparisons of the sizes of the compiled forms obtained using C2D, when no preprocessing is used vs. when the eq combination has been applied first.

## Impact of eq on Dsharp (compilation times)



Figure: Comparisons of the compilation times of Dsharp, when no preprocessing is used vs. when the eq combination has been applied first.

## Impact of eq on Dsharp (sizes of the compiled forms)



Figure: Comparisons of the sizes of the compiled forms obtained using Dsharp, when no preprocessing is used vs. when the eq combination has been applied first.

## Empirical Results

- Empirically, each of eq and \#eq proves to be a useful preprocessing combination (whatever the downstream model counter)
- The gate-detection-and-replacement preprocessings appear as particularly interesting for improving search-based model counters
- However this family of preprocessings is restricted to a small subset of target gates
- Is it possible to do better, and to enlarge this family?


## Overview

## Introduction

## Reducing CNF Formulae <br> P-Preprocessings <br> NP-Preprocessings <br> Combining Preprocessings

Implicit GDR thanks to Definability

## Limitations of the Gate Detection and Replacement Preprocessings

- The replacement phase requires gates to be detected
- The search space for gates is huge
- The size of a gate can be huge as well


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- The replacement phase requires gates to be detected
- The search space for gates is huge
- The size of a gate can be huge as well
- Identifying "complex gates" is incompatible with the efficiency expected for a preprocessing: only "simple" gates are targeted

```
literal equivalences }\quady\leftrightarrow\mp@subsup{x}{1}{
AND/OR gates }\quady\leftrightarrow(\mp@subsup{x}{1}{}\wedge\overline{\mp@subsup{x}{2}{}}\wedge\mp@subsup{x}{3}{}
XOR gates }\quady\leftrightarrow(\mp@subsup{x}{1}{}\oplus\overline{\mp@subsup{x}{2}{}}
```


## Overcoming the Limitations (1)

- The (explicit) detection phase can be replaced by an implicit detection phase
- There is no need to identify $f$ to determine that a gate of the form $y \leftrightarrow f\left(x_{1}, \ldots, x_{n}\right)$ exists in $\Sigma$


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- Let us ask Evert and Alessandro for some help ...


## Evert Willem Beth (1908-1964)



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- $\Sigma$ explicitly defines $y$ in terms of $X=\left\{x_{1}, \ldots, x_{n}\right\}$ iff there exists a
 formula $f\left(x_{1}, \ldots, x_{n}\right)$ over $X$ such that

$$
\Sigma \models y \leftrightarrow f\left(x_{1}, \ldots, x_{n}\right)
$$

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$$
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$$

- $\Sigma$ implicitly defines $y$ in terms of $X=\left\{x_{1}, \ldots, x_{n}\right\}$ iff for every canonical term $\gamma_{X}$ over $X$, we have $\Sigma \wedge \gamma_{X} \vDash y$ or $\Sigma \wedge \gamma_{X} \vDash \bar{y}$


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$$
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$$

- $\Sigma$ implicitly defines $y$ in terms of $X=\left\{x_{1}, \ldots, x_{n}\right\}$ iff for every canonical term $\gamma_{X}$ over $X$, we have $\Sigma \wedge \gamma_{X} \vDash y$ or $\Sigma \wedge \gamma_{X} \vDash \bar{y}$
- Beth's theorem: $\Sigma$ explicitly defines $y$ in terms of $X$ iff $\Sigma$ implicitly defines $y$ in terms of $X$


## Alessandro Padoa (1868-1937)



Padoa's theorem:

Let $\Sigma_{X}^{\prime}$ be equal to $\Sigma$ where each variable but those of $X$ have been renamed in a uniform way
If $y \notin X$, then $\Sigma$ (implicitly) defines $y$ in terms of $X$ iff $\Sigma \wedge \Sigma_{X}^{\prime} \wedge y \wedge \overline{y^{\prime}}$ is inconsistent

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inconsistent

Deciding whether $\Sigma$ (implicitly) defines $y$ in terms of $X$ is "only" coNP-complete

## Overcoming the Limitations (2)

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- Explicit definability = Implicit definability (Beth's theorem)


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- There is no need to identify $f$ to determine that a gate of the form $y \leftrightarrow f\left(x_{1}, \ldots, x_{n}\right)$ exists in $\Sigma$
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- One call to a SAT solver is enough to decide whether $\Sigma$ defines $y$ in terms of $\left\{x_{1}, \ldots, x_{n}\right\}$ (thanks to Padoa's theorem)


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- One call to a SAT solver is enough to decide whether $\Sigma$ defines $y$ in terms of $\left\{x_{1}, \ldots, x_{n}\right\}$ (thanks to Padoa's theorem)
- There is no need to identify $f$ to compute $\Sigma\left[y \leftarrow f\left(x_{1}, \ldots, x_{n}\right)\right]$


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- One call to a SAT solver is enough to decide whether $\Sigma$ defines $y$ in terms of $\left\{x_{1}, \ldots, x_{n}\right\}$ (thanks to Padoa's theorem)
- There is no need to identify $f$ to compute $\Sigma\left[y \leftarrow f\left(x_{1}, \ldots, x_{n}\right)\right]$
- The replacement phase can be replaced by an output variable elimination phase: if $y \leftrightarrow f\left(x_{1}, \ldots, x_{n}\right)$ is a gate of $\Sigma$, then

$$
\Sigma\left[y \leftarrow f\left(x_{1}, \ldots, x_{n}\right)\right] \equiv \exists y . \Sigma
$$

## The B + E Preprocessor

A two-step preprocessing

- "Detection = Bipartition": compute a definability bipartition $\langle I, O\rangle$ of $\Sigma$ such that $I \cup O=\operatorname{Var}(\Sigma), I \cap O=\emptyset$, and $\Sigma$ defines every variable $o \in O$ in terms of $I$


## The B + E Preprocessor

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- "Replacement = Elimination":
compute $\exists E . \Sigma$ for $E \subseteq O$


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- "Replacement = Elimination":
compute $\exists E . \Sigma$ for $E \subseteq O$
- Steps B and E of B + E can be tuned in order to keep the preprocessing phase light from a computational standpoint


## Back to GDR

$$
\Sigma=\begin{aligned}
& \bar{x} \vee u \vee v \\
& \\
& \bar{x} \vee \bar{y} \vee u \\
& \bar{x} \vee \bar{z} \vee u \\
& \\
& x \vee \bar{u} \\
& \\
& y \vee z \vee \bar{u}
\end{aligned} \quad u \leftrightarrow(x \wedge(y \vee z))
$$

$\Sigma$ defines $u$ in terms of $\{x, y, z\}$

## Detecting $u$ as an Output Variable and Eliminating it

## Identification:

$\Sigma \wedge \Sigma_{\{x, y, z\}}^{\prime} \wedge u \wedge \overline{u^{\prime}}$ is inconsistent

$$
\begin{aligned}
& \bar{x} \vee u \vee v \\
& \bar{x} \vee \bar{y} \vee u \\
& \bar{x} \vee \bar{z} \vee u \\
& x \vee \bar{u} \\
& y \vee z \vee \bar{u} \\
& \bar{x} \vee u^{\prime} \vee v^{\prime} \\
& \bar{x} \vee \bar{y} \vee u^{\prime} \\
& \bar{x} \vee \bar{z} \vee u^{\prime} \\
& x \vee \overline{u^{\prime}} \\
& y \vee z \vee \overline{u^{\prime}} \\
& \frac{u}{u^{\prime}}
\end{aligned}
$$

## Detecting $u$ as an Output Variable and Eliminating it

## Identification:

$\Sigma \wedge \Sigma_{\{x, y, z\}}^{\prime} \wedge u \wedge \overline{u^{\prime}}$ is inconsistent

$$
\begin{aligned}
& \bar{x} \vee u \vee v \\
& \bar{x} \vee \bar{y} \vee u \\
& \bar{x} \vee \overline{\bar{z}} \vee u \\
& x \vee \bar{u} \\
& y \vee z \vee \bar{u} \\
& \bar{x} \vee u^{\prime} \vee v^{\prime} \\
& \bar{x} \vee \bar{y} \vee u^{\prime} \\
& \bar{x} \vee \bar{z} \vee u^{\prime} \\
& x \vee \overline{u^{\prime}} \\
& y \vee z \overline{u^{\prime}} \\
& \frac{u}{u^{\prime}}
\end{aligned}
$$

## Elimination:

computing resolvents over $u$

| $\bar{x} \vee v \vee x$ | valid |
| :--- | ---: |
| $\bar{x} \vee v \vee y \vee z$ |  |
| $\bar{x} \vee \bar{y} \vee x$ | valid |
| $\bar{x} \vee \bar{y} \vee y \vee z$ | valid |
| $\bar{x} \vee \bar{z} \vee x$ | valid |
| $\bar{x} \vee \bar{z} \vee y \vee z$ | valid |

## Detecting $u$ as an Output Variable and Eliminating it

## Identification:

$\Sigma \wedge \Sigma_{\{x, y, z\}}^{\prime} \wedge u \wedge \overline{u^{\prime}}$ is inconsistent

$$
\begin{aligned}
& \bar{x} \vee u \vee v \\
& \bar{x} \vee \bar{y} \vee u \\
& \bar{x} \vee \bar{z} \vee u \\
& x \vee \bar{u} \\
& y \vee z \vee \bar{u} \\
& \bar{x} \vee u^{\prime} \vee v^{\prime} \\
& \bar{x} \vee \bar{y} \vee u^{\prime} \\
& \bar{x} \vee \bar{z} \vee u^{\prime} \\
& x \vee \overline{u^{\prime}} \\
& y \vee z \vee \overline{u^{\prime}} \\
& u \\
& \overline{u^{\prime}}
\end{aligned}
$$

## Elimination:

computing resolvents over $u$

| $\bar{x} \vee v \vee x$ | valid |
| :--- | ---: |
| $\bar{x} \vee v \vee y \vee z$ |  |
| $\bar{x} \vee \bar{y} \vee x$ | valid |
| $\bar{x} \vee \bar{y} \vee y \vee z$ | valid |
| $\bar{x} \vee \bar{z} \vee x$ | valid |
| $\bar{x} \vee \bar{z} \vee y \vee z$ | valid |

$$
\|\Sigma\|=\|\bar{x} \vee v \vee y \vee z\|=15
$$

## Tuning the Computational Effort

Both steps B and E of B + E can be tuned in order to keep the preprocessing phase light from a computational standpoint

- It is not necessary to determine a definability bipartition $\langle I, O\rangle$ with |I| minimal
$\Rightarrow B$ is a greedy algorithm (one definability test per variable)
$\Rightarrow$ Only the minimality of $I$ for $\subseteq$ is guaranteed


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- It is not necessary to determine a definability bipartition $\langle I, O\rangle$ with |I| minimal
$\Rightarrow B$ is a greedy algorithm (one definability test per variable)
$\Rightarrow$ Only the minimality of $I$ for $\subseteq$ is guaranteed
- It is not necessary to eliminate in $\Sigma$ every variable of $O$ but focusing on a subset $E \subseteq O$ is enough
$\Rightarrow$ Eliminating every output variable could lead to an exponential blow up
$\Rightarrow$ The elimination of $y \in O$ is committed only if $|\Sigma|$ after the elimination step and some additional preprocessing (occurrence simplification and vivification) remains small enough


## Experiments

## Objectives:

- Evaluating the computational benefits offered by B + E when used upstream to state-of-the-art model counters:
- the search-based model counter Cachet
- the search-based model counter SharpSAT
- the compilation-based model counter C2D


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- Evaluating the computational benefits offered by B + E when used upstream to state-of-the-art model counters:
- the search-based model counter Cachet
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- the compilation-based model counter C2D
- Comparing the benefits offered by $B+E$ with those offered by our previous preprocessor pmc (based on gate identification and replacement) or with no preprocessing


## Empirical Results



## Empirical Results

$B+E$ vs. no preprocessing


Figure: Model counting time reductions achieved by $B+E$ vs. no preprocessing

## Empirical Results

$B+E$ vs. pmc

(a) $\mathrm{B}+\mathrm{E}+$ Cachet vs. pmc+Cachet

(b) $\mathrm{B}+\mathrm{E}+\mathrm{C} 2 \mathrm{D}$ vs. $\mathrm{pmc}+\mathrm{C} 2 \mathrm{D}$

Figure: Model counting time reductions achieved by B + Evs. pmc

## Empirical Results

- The experiments clearly show the benefits offered by B+E
- B + E appears typically as a better preprocessor than pmc since it leads typically to improved performances (smaller computation times)


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## NP-Preprocessing

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