#### **NP-Preprocessing**

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Research School on Knowledge Compilation, ENS Lyon, December 4th-8th, 2017

#### Introduction

Reducing CNF Formulae

**P-Preprocessings** 

**NP-Preprocessings** 

Combining Preprocessings

Implicit GDR thanks to Definability

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Two **preprocessing** approaches for circumventing the complexity of computationally hard tasks

knowledge compilation

input:  $\Sigma$  (in  $\mathcal{L}_1$ ) — compilation  $\rightarrow \Psi$  (in  $\mathcal{L}_2$ ) — resolution  $\rightarrow$  output: result input:  $\alpha$  (in  $\mathcal{L}_1$ ) — Two **preprocessing** approaches for circumventing the complexity of computationally hard tasks

► knowledge compilation input:  $\Sigma$  (in  $\mathcal{L}_1$ ) — compilation →  $\Psi$  (in  $\mathcal{L}_2$ ) — resolution → output: result input:  $\alpha$  (in  $\mathcal{L}_1$ ) —

► preprocessing input:  $\Sigma, \alpha$  (in  $\mathcal{L}_1$ ) - preprocessing  $\rightarrow \Psi$  (in  $\mathcal{L}_1$ ) - resolution  $\rightarrow$  output: result Two **preprocessing** approaches for circumventing the complexity of computationally hard tasks

 knowledge compilation input: Σ (in L<sub>1</sub>) \_\_\_\_\_ compilation → Ψ (in L<sub>2</sub>) \_\_\_\_\_ resolution → output: result input: α (in L<sub>1</sub>) \_\_\_\_\_
 preprocessing input: Σ, α (in L<sub>1</sub>) \_\_\_\_\_ resolution → output: result

#### 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 α of the input (does not exist in general, can be preprocessed as well or not)

#### SAT

- Input: a CNF formula Σ
- Output: 1 if Σ 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!

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

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• 
$$\Sigma = (x \lor y) \land (\neg y \lor z)$$

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

#### $\blacktriangleright \Sigma \mapsto \|\Sigma\| = ?$

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• The models of  $\Sigma$  over  $\{x, y, z\}$  are :

 $\blacktriangleright \Sigma \mapsto \|\Sigma\| = ?$ 

$$\blacktriangleright \Sigma = (x \lor y) \land (\neg y \lor z)$$

• The models of  $\Sigma$  over  $\{x, y, z\}$  are :

$$\bullet \|\Sigma\| = 4$$

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

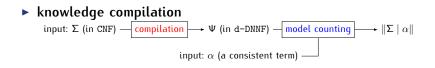
Seinosuke Toda (Gödel Prize 1998):

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

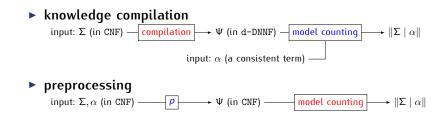
#### Many model counters have been developed:

- 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 vs. Preprocessing for Model Counting



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## **P-Preprocessings**

- Polynomial-time preprocessings
- Can prove useful for SAT and #SAT
- ▶ Related to the notion of kernelization: given a parameterized problem  $(L \subseteq V^*, \kappa : V^* \to \mathbb{N})$ , 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 κ(.) 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)

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

►

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SAT solvers and model counters typically considers CNF inputs

- ► The CNF assumption is not restrictive
- $\blacktriangleright$  Every circuit  $\Psi$  can be turned into a CNF formula  $\Sigma$  in linear time
- The translation requires the introduction of new variables  $Y = \{y_1, \ldots, \}$  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)

$$\begin{split} \Psi &= (x_1 \wedge \bar{x}_2) \lor (x_2 \wedge x_3) \\ T(\Psi) &\equiv (y_1 \lor y_2) \land (y_1 \Leftrightarrow (x_1 \wedge \bar{x}_2)) \land (y_2 \Leftrightarrow (x_2 \wedge x_3)) \\ T(\Psi) &= \\ y_1 \lor y_2 \\ \bar{y}_1 \lor x_1 \\ \bar{y}_1 \lor \bar{x}_2 \\ y_1 \lor \bar{x}_1 \lor x_2 \\ \bar{y}_2 \lor x_2 \\ \bar{y}_2 \lor x_3 \\ y_2 \lor \bar{x}_2 \lor \bar{x}_3 \end{split}$$

$$\begin{split} \Psi &= (x_1 \wedge \bar{x}_2) \lor (x_2 \wedge x_3) \\ \mathcal{P}G(\Psi) &\equiv (y_1 \lor y_2) \land (y_1 \Rightarrow (x_1 \wedge \bar{x}_2)) \land (y_2 \Rightarrow (x_2 \wedge x_3)) \end{split}$$

 $\begin{array}{l} \mathsf{PG}(\Psi) = \\ y_1 \lor y_2 \\ \bar{y}_1 \lor x_1 \\ \bar{y}_1 \lor \bar{x}_2 \\ \bar{y}_2 \lor x_2 \\ \bar{y}_2 \lor x_3 \\ \bar{y}_2 \lor x_3 \end{array}$ 

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#### $\texttt{CNF}\ \Sigma\mapsto\texttt{CNF}\ \rho(\Sigma)$

- What are the connections between  $\Sigma$  and  $p(\Sigma)$ ?
- Removing clauses from Σ
- Removing literals in the clauses of  $\Sigma$
- Updating Σ 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  $\Sigma \setminus {\delta} \models \delta$
- A CNF Σ is irredundant if and only if it does not contain any redundant clause
- A subset Σ' of a CNF Σ is an irredundant equivalent subset (IES) of Σ if and only if Σ' is irredundant and Σ' ≡ Σ
- Deciding whether a CNF  $\Sigma$  is irredundant is NP-complete
- Deciding whether a CNF Σ' is an irredundant equivalent subset (IES) of a CNF Σ is D<sup>p</sup>-complete
- Given an integer k, deciding whether a CNF Σ has an IES of size at most k is Σ<sub>2</sub><sup>p</sup>-complete
- Given an integer k, deciding whether there exists a CNF formula  $\Sigma'$  with at most k literals (or with at most k clauses) equivalent to a given CNF  $\Sigma$  is  $\Sigma_2^p$ -complete

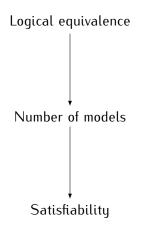
Redundancy can be helpful (favoring the unit propagation power by adding empowering clauses)

- $x_3 \lor x_4$  is a logical consequence of
  - $\Sigma = x_1 \lor x_2 \lor x_3$  $\bar{x}_1 \lor x_2 \lor x_4$  $x_1 \lor \bar{x}_2 \lor x_3$  $\bar{x}_1 \lor \bar{x}_2 \lor x_4$
- Assuming  $\bar{x}_3 \wedge \bar{x}_4$ , there is no unit refutation from  $\Sigma$
- ► If  $x_3 \lor x_4$  (or  $x_2 \lor x_3 \lor x_4$  or  $\overline{x}_2 \lor x_3 \lor x_4$ ) is added to  $\Sigma$ , a unit refutation from  $\Sigma$  under the assumption  $\overline{x}_3 \land \overline{x}_4$  exists
- ► Learning clauses is a key ingredient for efficient SAT solvers based on the CDCL architecture!

- Logical equivalence
- Number of models
- Satisfiability

### Levels of Preservation

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



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

#### Several measures for the reduction achieved can be considered:

- The number of variables in the input CNF  $\Sigma$
- The size of Σ (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  $\delta_2$ 

 $SE: (x_1 \lor x_2) \land (x_1 \lor x_2 \lor \bar{x}_3) \mapsto x_1 \lor x_2$ 

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

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

 $PLE: (x_1 \lor x_2 \lor x_3) \land (\bar{x}_2 \lor \bar{x}_3) \mapsto \bar{x}_2 \lor \bar{x}_3$ 

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

## 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 =1h
- Memory-out = 7.6 GiB

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P-Preprocessings Occurrence Reduction (OR) Vivification (VI) Gate Detection and Replacement (GDR)

**NP-Preprocessings** 

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- Occurrence reduction aims to remove some literals in the clauses of Σ
- ► 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  $\Sigma \land \ell_{j+1} \land \sim \ell_1 \land \ldots \land \sim \ell_j$  is contradictory
- When this is the case, ℓ<sub>j+1</sub> can be removed from α without questioning logical equivalence
- bcp is used as an incomplete yet efficient method to solve the entailment problem

Algorithm 1: OR Occurrence reduction

input : a CNF formula  $\boldsymbol{\Sigma}$ 

output: a CNF formula equivalent to  $\boldsymbol{\Sigma}$ 

1 
$$\mathcal{L} \leftarrow sort(Lit(\Sigma))$$

2 foreach  $\ell \in \mathcal{L}$  do

3 foreach 
$$\alpha \in \Sigma$$
 s.t.  $\ell \in \alpha$  do

4 
$$\begin{bmatrix} \text{if } \emptyset \in \text{bcp}(\Sigma \cup \{\ell\} \cup \{\sim(\alpha \setminus \{\ell\})\}) \text{ then} \\ \Sigma \leftarrow (\Sigma \setminus \{\alpha\}) \cup \{\alpha \setminus \{\ell\}\}; \end{bmatrix}$$

5 return Σ

## Occurrence Reduction (OR): Example

$$\Sigma = 
 a \lor f \qquad a \lor b \lor c 
 b \lor d \lor e \qquad c \lor \neg d \lor e 
 b \lor d \lor \neg e \qquad c \lor \neg d \lor \neg e$$

$$\mathcal{L} = (b, c, e, \neg e, d, \neg d, a, f, \neg a, \neg b, \neg c)$$

$$OR(\Sigma) = 
 a \lor f \qquad b \lor c 
 b \lor d \qquad c \lor \neg d$$

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

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- Preserves logical equivalence
- Neither is confluent nor is projective
- $\#var(\Sigma) \geq \#var(OR(\Sigma))$
- $\#lit(\Sigma) \geq \#lit(OR(\Sigma))$
- $tw(\Sigma) \ge tw(OR(\Sigma))$

### **OR:** Reduction of the Number of Variables

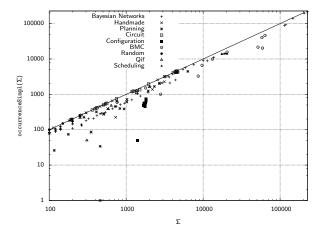


FIGURE: Comparing  $\#var(\Sigma)$  with  $\#var(OR(\Sigma))$ .

### OR: Reduction of the Size

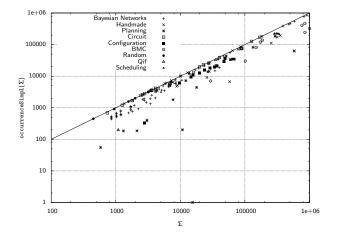
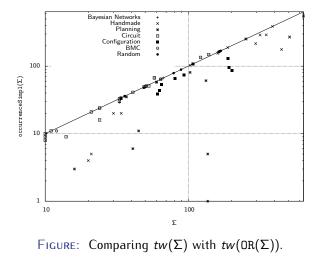


FIGURE: Comparing  $\#lit(\Sigma)$  with  $\#lit(OR(\Sigma))$ .

### OR: Reduction of the Treewidth



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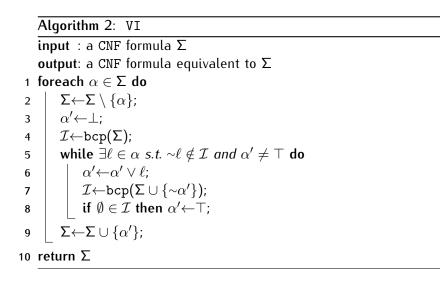
**NP-Preprocessings** 

Combining Preprocessings

Implicit GDR thanks to Definability

# Vivification (VI)

- Vivification aims to reduce Σ, i.e., to remove some clauses and some literals in Σ while preserving equivalence
- Given a clause  $\alpha = \ell_1 \lor \ldots \lor \ell_k$  of  $\Sigma$  two rules are used in order to determine whether  $\alpha$  can be removed from  $\Sigma$  or simply shortened
- Let  $\alpha'$  be any subclause of  $\alpha$
- On the one hand, if for any j ∈ 0,..., k − 1, one can prove using bcp that Σ \ {α} ⊨ α', then for sure α is entailed by Σ \ {α} so that α can be removed from Σ
- On the other hand, if one can prove using bcp that  $\Sigma \setminus \{\alpha\} \models \alpha' \lor \sim \ell_{j+1}$ , then  $\ell_{j+1}$  can be removed from  $\alpha$  without questioning equivalence



$$\Sigma = \\ a \lor b \lor c \lor d \\ a \lor b \lor c \\ a \lor \neg d$$

Assume that the variables are processed w.r.t. the ordering d < c < b < aVI( $\Sigma$ ) =  $a \lor b \lor c$ 

$$a \vee \neg d$$

The effect of VI on  $\Sigma$  is to eliminate the first clause  $a \lor b \lor c \lor d$ 

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- Preserves logical equivalence
- Neither is confluent nor is projective
- $\#var(\Sigma) \geq \#var(VI(\Sigma))$
- $\#lit(\Sigma) \geq \#lit(VI(\Sigma))$
- $tw(\Sigma) \ge tw(VI(\Sigma))$

### VI: Reduction of the Number of Variables

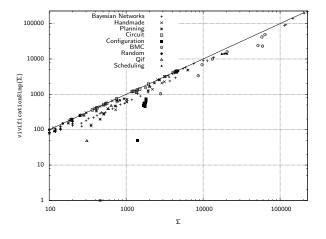


FIGURE: Comparing  $\#var(\Sigma)$  with  $\#var(VI(\Sigma))$ .

### VI: Reduction of the Size

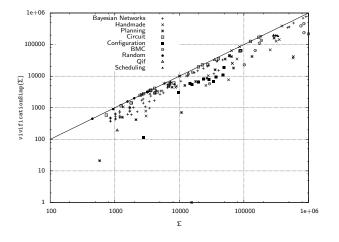


FIGURE: Comparing  $\#lit(\Sigma)$  with  $\#lit(VI(\Sigma))$ .

## VI: Reduction of the Treewidth

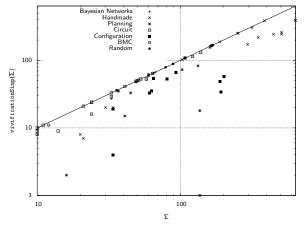


FIGURE: Comparing  $tw(\Sigma)$  with  $tw(VI(\Sigma))$ .

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#### P-Preprocessings

#### Occurrence Reduction (OR)

#### Vivification (VI)

#### Gate Detection and Replacement (GDR)

Literal Equivalence (LE) AND/OR Gate Equivalence (AG) XOR Gate Equivalence (XG)

#### NP-Preprocessings

#### Combining Preprocessings

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)

$$\Sigma = \frac{\overline{x} \lor u \lor v}{\overline{x} \lor \overline{y} \lor u}$$
$$\Sigma = \frac{\overline{x} \lor \overline{z} \lor u}{x \lor \overline{z} \lor u}$$
$$y \lor z \lor \overline{u}$$

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$$\Sigma = \begin{array}{ccc} \overline{x} \lor u \lor v \\ \overline{x} \lor \overline{y} \lor u \\ \overline{x} \lor \overline{z} \lor u \\ x \lor \overline{u} \\ y \lor z \lor \overline{u} \end{array} \qquad \qquad u \leftrightarrow (x \land (y \lor z))$$

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 $\Sigma \equiv (\overline{x} \lor u \lor v) \land (u \leftrightarrow (x \land (y \lor z)))$  detection

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$$\begin{split} \Sigma &\equiv \\ & (\overline{x} \lor u \lor v) \land (u \leftrightarrow (x \land (y \lor z))) & \text{detection} \\ & (\overline{x} \lor (x \land (y \lor z)) \lor v) \land (u \leftrightarrow (x \land (y \lor z))) & \text{replacement} \end{split}$$

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)

 $\Sigma = \begin{array}{c} \overline{x} \lor u \lor v \\ \overline{x} \lor \overline{y} \lor u \\ \overline{x} \lor \overline{z} \lor u \\ x \lor \overline{u} \\ y \lor z \lor \overline{u} \end{array} \qquad \qquad u \leftrightarrow (x \land (y \lor z))$ 

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$$\begin{split} \Sigma &\equiv \\ & (\overline{x} \lor u \lor v) \land (u \leftrightarrow (x \land (y \lor z))) & \text{detection} \\ & (\overline{x} \lor (x \land (y \lor z)) \lor v) \land (u \leftrightarrow (x \land (y \lor z))) & \text{replacement} \\ & (\overline{x} \lor y \lor z \lor v) \land (u \leftrightarrow (x \land (y \lor z))) & \text{normalization} \end{split}$$

$$\|\Sigma\| = \|\Sigma[u \leftarrow (x \land (y \lor z))]\| = \|\overline{x} \lor y \lor z \lor v\| = 15$$

- 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 *l*, all the literals *l'* which can be found equivalent to *l* using bcp are replaced by *l* in Σ
- ► Taking advantage of bcp makes it more efficient than a "syntactic detection" (if two binary clauses stating an equivalence between two literals ℓ and ℓ' occur in Σ, then those literals are found equivalent using bcp, but the converse does not hold)

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  $\exists \ell \in Lit(\Phi)$  s.t.  $var(\ell)$  is not marked do // detection mark  $var(\ell)$ ; 3  $\mathcal{P}_{\ell} \leftarrow bcp(\Phi \cup \{\ell\});$ 4  $\mathcal{N}_{\ell} \leftarrow bcp(\Phi \cup \{\sim \ell\});$ 5  $\Gamma \leftarrow \{\ell \leftrightarrow \ell' | \ell' \neq \ell \text{ and } \ell' \in \mathcal{P}_{\ell} \text{ and } \sim \ell' \in \mathcal{N}_{\ell}\};$ 6 // replacement foreach  $\ell \leftrightarrow \ell' \in \Gamma$  do 7 replace  $\ell$  by  $\ell'$  in  $\Phi$ ; 8 9 return Φ

# Literal Equivalence (LE): Example

$$\Sigma = a \lor b \lor c \lor \neg d \quad \neg a \lor \neg b \lor \neg c \lor d a \lor b \lor \neg c \quad \neg a \lor \neg b \lor c \neg a \lor b \quad a \lor \neg b \lor c \neg e \lor \neg f \lor h \quad e \lor f \lor g e \lor \neg g \quad \neg e \lor \neg h$$

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

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

$$LE(\Sigma) = \\ \neg f \lor \neg g \quad f \lor \neg h$$

- Preserves the number of models (but not logical equivalence)
- Neither is confluent nor is projective
- $\#var(\Sigma) \geq \#var(LE(\Sigma))$
- $\#lit(\Sigma) \geq \#lit(LE(\Sigma))$
- $tw(\Sigma) \not\geq tw(LE(\Sigma))$

### LE: Reduction of the Number of Variables

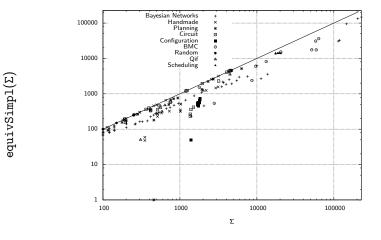


FIGURE: Comparing  $\#var(\Sigma)$  with  $\#var(LE(\Sigma))$ .

### LE: Reduction of the Size

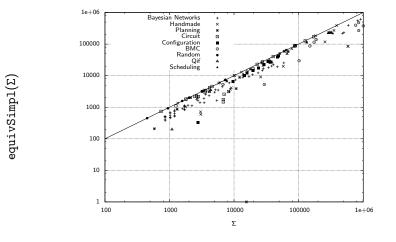


FIGURE: Comparing  $\#lit(\Sigma)$  with  $\#lit(LE(\Sigma))$ .

### LE: Reduction of the Treewidth

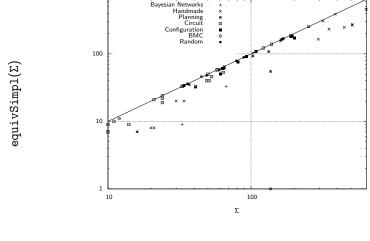


FIGURE: Comparing  $tw(\Sigma)$  with  $tw(LE(\Sigma))$ .

# 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)

#### Algorithm 4: AG

```
input
                                   : a CNF formula \Sigma
          output
                                   : a CNF formula \Phi such that \|\Phi\| = \|\Sigma\|
          \Phi \leftarrow \Sigma;
          // detection
    2
          \Gamma \leftarrow \emptyset; Unmark all literals of \Phi;
   34567
          while \exists \ell \in Lit(\Phi) s.t. \ell is not marked do
                       mark ℓ:
                       \mathcal{P}_{\ell} \leftarrow (\operatorname{bcp}(\Phi \cup \{\ell\}) \setminus (\operatorname{bcp}(\Phi) \cup \{\ell\})) \cup \{\sim \ell\};
                       if \emptyset \in bcp(\Phi \cup \mathcal{P}_{\ell}) then
                                    let \mathcal{C}_{\ell} \subseteq \mathcal{P}_{\ell} s.t. \emptyset \in bcp(\Phi \cup \mathcal{C}_{\ell}) and \sim \ell \in \mathcal{C}_{\ell};
                                    \Gamma \leftarrow \Gamma \cup \{\ell \leftrightarrow \bigwedge_{\ell' \in \mathcal{C}_{\ell} \setminus \{\sim_{\ell}\}} \ell'\};
   8
          // replacement
   9
          while \exists \ell \leftrightarrow \beta \in \Gamma st. |\beta| < maxA and |\Phi[\ell \leftarrow \beta]| < |\Phi| do
10
                       \Phi \leftarrow \Phi[\ell \leftarrow \beta];
11
                      \Gamma \leftarrow \Gamma[\ell \leftarrow \beta]:
12 return Φ
```

# AND/OR Gate Equivalence (AG): Example

$$\Sigma = a \lor b \lor c \lor d \quad \neg a \lor \neg b \neg a \lor b \lor \neg c \quad \neg a \lor b \lor c \lor \neg d \neg a \lor e \qquad a \lor f$$

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

$$AG(\Sigma) = b \lor c \lor d \lor e \quad \neg b \lor f \\ \neg c \lor f \quad \neg d \lor f$$

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- Preserves the number of models (but not logical equivalence)
- Neither is confluent nor is projective
- $\#var(\Sigma) \geq \#var(AG(\Sigma))$
- $\#lit(\Sigma) \geq \#lit(AG(\Sigma))$
- $tw(\Sigma) \not\geq tw(AG(\Sigma))$

#### AG: Reduction of the Number of Variables

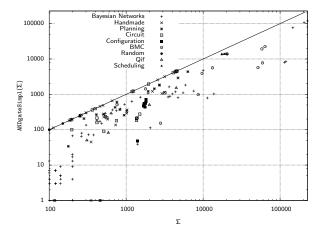


FIGURE: Comparing  $\#var(\Sigma)$  with  $\#var(AG(\Sigma))$ .

## AG: Reduction of the Size

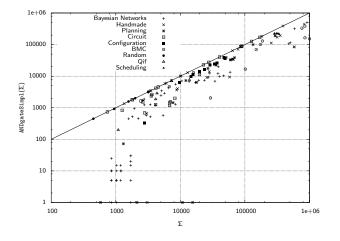


FIGURE: Comparing  $\#lit(\Sigma)$  with  $\#lit(AG(\Sigma))$ .

## AG: Reduction of the Treewidth

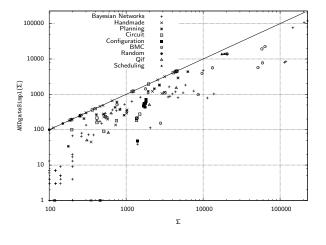


FIGURE: Comparing  $tw(\Sigma)$  with  $tw(AG(\Sigma))$ .

# 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

Algorithm 5: XG

input : a CNF formula  $\boldsymbol{\Sigma}$ 

**output**: a CNF formula  $\Phi$  such that  $\|\Phi\| = \|\Sigma\|$ 

- 1 Φ←Σ;
  - // detection

2  $\Gamma \leftarrow \{XOR \ clauses \ syntactically \ detected\};$ 

// Gaussian elimination

- $3 \ \mathsf{\Gamma} \leftarrow \mathsf{Gauss}(\{\ell_1 \leftrightarrow \chi_1, \ell_2 \leftrightarrow \chi_2, \dots, \ell_k \leftrightarrow \chi_k\})$ 
  - // replacement
- 4 for  $i \leftarrow 1$  to k do

5 
$$\lfloor$$
 if  $\nexists \alpha \in \Phi[\ell_i \leftarrow \chi_i]$  s.t.  $|\alpha| > maxX$  then  $\Phi \leftarrow \Phi[\ell_i \leftarrow \chi_i]$ ;

6 return Φ

# XOR Gate Equivalence (XG): Example

$$\Sigma = \\ b \lor d \qquad \neg b \lor \neg d \\ \neg a \lor \neg b \lor c \qquad a \lor \neg b \lor \neg c \\ \neg a \lor b \lor \neg c \qquad a \lor b \lor c \\ b \lor e \qquad a \lor f$$

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

$$XG(\Sigma) = \neg d \lor e \qquad c \lor d \lor f \neg c \lor \neg d \lor f$$

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

- Preserves the number of models (but not logical equivalence)
- Neither is confluent nor is projective
- $\#var(\Sigma) \geq \#var(XG(\Sigma))$
- $\#lit(\Sigma) \geq \#lit(XG(\Sigma))$
- $tw(\Sigma) \not\geq tw(XG(\Sigma))$

#### XG: Reduction of the Number of Variables

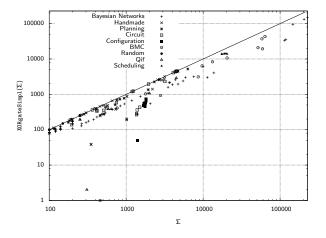


FIGURE: Comparing  $\#var(\Sigma)$  with  $\#var(XG(\Sigma))$ .

## XG: Reduction of the Size

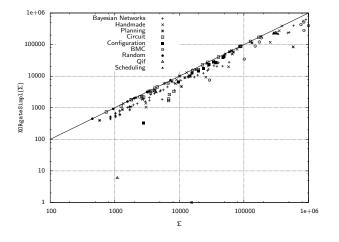


FIGURE: Comparing  $\#lit(\Sigma)$  with  $\#lit(XG(\Sigma))$ .

## XG: Reduction of the Treewidth

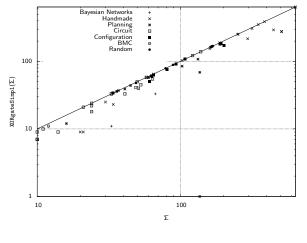


FIGURE: Comparing  $tw(\Sigma)$  with  $tw(XG(\Sigma))$ .

Introduction

Reducing CNF Formulae

**P-Preprocessings** 

#### NP-Preprocessings Backbone Identification (BI)

Combining Preprocessings

Implicit GDR thanks to Definability

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- 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

Introduction

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#### NP-Preprocessings Backbone Identification (BI)

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- 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 BI preprocessing is to make the backbone B of the input CNF formula Σ explicit, to conjoin it to Σ, and to use bcp (Boolean Constraint Propagation) on the resulting set of clauses

Algorithm 6: BI Backbone Identification

input : a CNF formula  $\boldsymbol{\Sigma}$ 

**output**: the CNF  $bcp(\Sigma \cup B)$ , where  $\mathcal{B}$  is the backbone of  $\Sigma$ 

1 
$$\mathcal{B} \leftarrow \emptyset;$$

$$\begin{array}{l} 2 \ \mathcal{I} \leftarrow \texttt{solve}(\boldsymbol{\Sigma}); \\ 3 \ \texttt{while} \ \exists \ell \in \mathcal{I} \ s.t. \ \ell \notin \mathcal{B} \ \texttt{do} \\ 4 \ \left[ \begin{array}{c} \mathcal{I}' \leftarrow \texttt{solve}(\boldsymbol{\Sigma} \cup \{\sim \ell\}); \\ \texttt{if} \ \mathcal{I}' = \emptyset \ \texttt{then} \ \mathcal{B} \leftarrow \mathcal{B} \cup \{\ell\} \texttt{else} \ \mathcal{I} \leftarrow \mathcal{I} \cap \mathcal{I}' \end{array} \right] \end{array}$$

6 return  $bcp(\Sigma \cup \mathcal{B})$ 

$$\Sigma = a \lor b$$

$$\neg a \lor b$$

$$\neg b \lor c$$

$$c \lor d$$

$$\neg c \lor e \lor f$$

$$f \lor \neg g$$

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

$$BI(\Sigma) = b$$

$$c$$

$$e \lor f$$

$$f \lor \neg g$$

- Preserves logical equivalence
- Is confluent and projective
- $\#var(\Sigma) \geq \#var(BI(\Sigma))$
- $\#lit(\Sigma) \geq \#lit(\mathtt{BI}(\Sigma))$
- $tw(\Sigma) \ge tw(BI(\Sigma))$

#### BI: Reduction of the Number of Variables

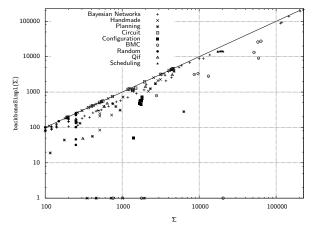


FIGURE: Comparing  $\#var(\Sigma)$  with  $\#var(BI(\Sigma))$ .

## BI: Reduction of the Size

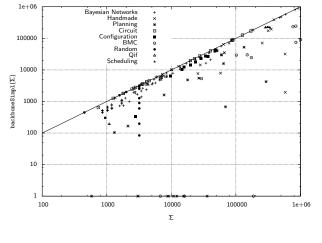


FIGURE: Comparing  $\#lit(\Sigma)$  with  $\#lit(BI(\Sigma))$ .

## BI: Reduction of the Treewidth

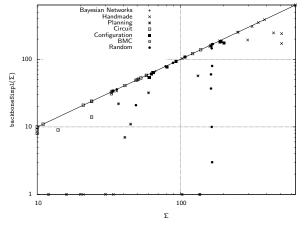


FIGURE: Comparing  $tw(\Sigma)$  with  $tw(BI(\Sigma))$ .

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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 optV = optB = optO = 1 and optG = 0. It is equivalence-preserving.
  - ▶ #eq corresponds to the parameter assignment of pmc where optV = optB = optO = 1 and optG = 1. This combination is guaranteed only to preserve the number of models of the input.

Algorithm 7: pmc
input : a CNF formula Σ
<b>output</b> : a CNF formula $\Phi$ such that $\ \Phi\  = \ \Sigma\ $
1 Φ←Σ;
2 if optB then $\Phi \leftarrow BI(\Phi)$ ;
3 <i>i</i> ←0;
4 while <i>i</i> < <i>numTries</i> do
5 $i \leftarrow i + 1;$
6 if optO then $\Phi \leftarrow OR(\Phi)$ ;
7 if $optG$ then $\Phi \leftarrow XG(AG(LE(\Phi)));$
8 if $optV$ then $\Phi \leftarrow VI(\Phi)$ ;
9 if fixpoint then break;
10 return Φ

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

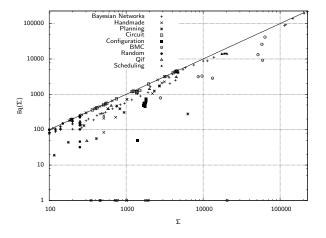


FIGURE: Comparing  $\#var(\Sigma)$  with  $\#var(eq(\Sigma))$ .

## eq: Reduction of the Size

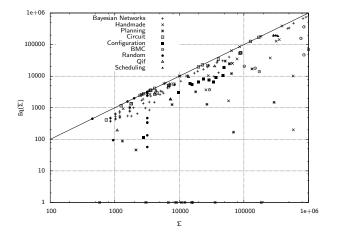


FIGURE: Comparing  $\#lit(\Sigma)$  with  $\#lit(eq(\Sigma))$ .

## eq: Reduction of the Treewidth

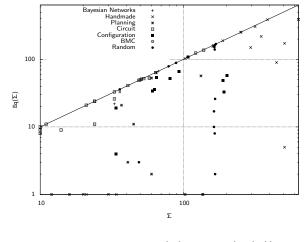


FIGURE: Comparing  $tw(\Sigma)$  with  $tw(eq(\Sigma))$ .

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

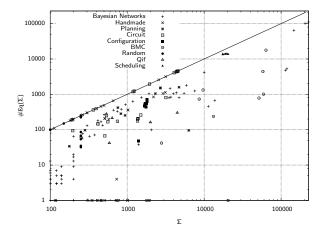


FIGURE: Comparing  $\#var(\Sigma)$  with  $\#var(\#eq(\Sigma))$ .

## #eq: Reduction of the Size

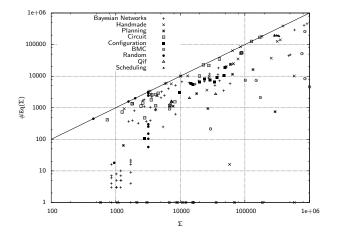


FIGURE: Comparing  $\#lit(\Sigma)$  with  $\#lit(\#eq(\Sigma))$ .

## #eq: Reduction of the Treewidth

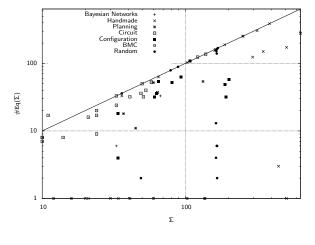


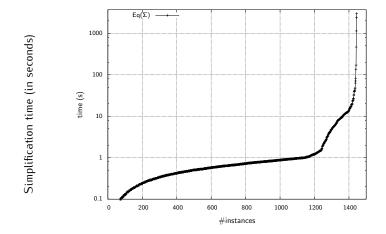
FIGURE: Comparing  $tw(\Sigma)$  with  $tw(\#eq(\Sigma))$ .

# Improving Model Counting?

- Reducing #var(Σ), #lit(Σ), and tw(Σ) 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)

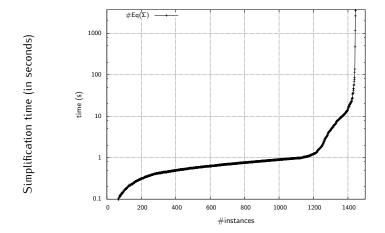
- 1449 CNF instances from the SAT LIBrary
- 9 data sets: BMC (18), Circuit (68), Qif (7), Planning (34), Random (105), Scheduling (6), Handmade (58), Configuration (35), Bayesian networks (1118)
- Cluster of Intel Xeon E5-2643 (3.30 GHz) processors with 32 GiB RAM on Linux CentOS
- Time-out =1h
- Memory-out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

# Efficiency of eq



Number of instances

# Efficiency of *#eq*



Number of instances

#### Impact of eq on Cachet

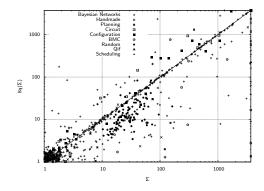


FIGURE: Comparison of the computation times needed to count the number of models of an instance using Cachet, when no preprocessing is used vs. when the *eq* combination has been applied first.

#### Impact of #eq on Cachet

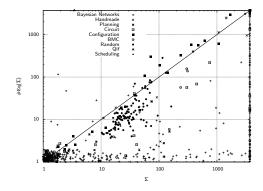


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#### Impact of eq on SharpSAT

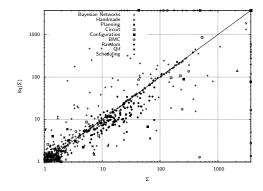


FIGURE: Comparison of the computation times needed to count the number of models of an instance using SharpSAT, when no preprocessing is used vs. when the *eq* combination has been applied first.

#### Impact of #eq on SharpSAT

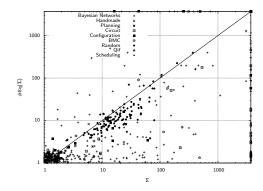


FIGURE: Comparison of the computation times needed to count the number of models of an instance using SharpSAT, when no preprocessing is used vs. when the #eq combination has been applied first.

#### Impact of eq on C2D (compilation times)

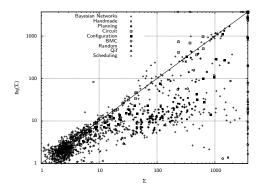


FIGURE: Comparisons of the compilation times of C2D, when no preprocessing is used vs. when the *eq* combination has been applied first.

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

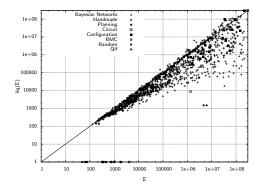


FIGURE: Comparisons of the sizes of the compiled forms obtained using C2D, when no preprocessing is used vs. when the *eq* combination has been applied first.

#### Impact of eq on Dsharp (compilation times)

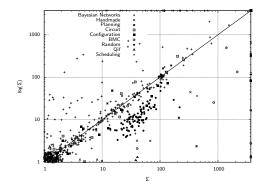


FIGURE: Comparisons of the compilation times of Dsharp, when no preprocessing is used vs. when the *eq* combination has been applied first.

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

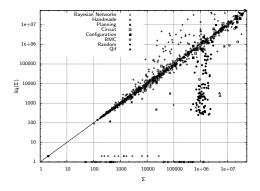


FIGURE: Comparisons of the sizes of the compiled forms obtained using Dsharp, when no preprocessing is used vs. when the *eq* combination has been applied first.

- 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?

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# 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

# 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
- Identifying "complex gates" is incompatible with the efficiency expected for a preprocessing: only "simple" gates are targeted

literal equivalences $y \leftrightarrow x_1$ AND/OR gates $y \leftrightarrow (x_1 \land \overline{x_2} \land x_3)$ XOR gates $y \leftrightarrow (x_1 \oplus \overline{x_2})$ 

- 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  $y \leftrightarrow f(x_1, ..., x_n)$  exists in  $\Sigma$

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- ► There is **no need to identify** f to determine that a gate of the form  $y \leftrightarrow f(x_1, ..., x_n)$  exists in  $\Sigma$
- ► Let us ask Evert and Alessandro for some help ...





•  $\Sigma$  explicitly defines *y* in terms of  $X = \{x_1, ..., x_n\}$  iff there exists a formula  $f(x_1, ..., x_n)$  over *X* such that

$$\Sigma \models y \leftrightarrow f(x_1, \ldots, x_n)$$



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►  $\Sigma$  implicitly defines y in terms of  $X = \{x_1, ..., x_n\}$  iff for every canonical term  $\gamma_X$  over X, we have  $\Sigma \land \gamma_X \models y$  or  $\Sigma \land \gamma_X \models \overline{y}$ 



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- Beth's theorem: Σ explicitly defines y in terms of X iff Σ implicitly defines y in terms of X

## Alessandro Padoa (1868-1937)



#### Padoa's theorem:

Let  $\Sigma'_X$  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  $\Sigma \wedge \Sigma'_X \wedge y \wedge \overline{y'}$  is inconsistent

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

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► There is **no need to identify** f to determine that a gate of the form  $y \leftrightarrow f(x_1, ..., x_n)$  exists in  $\Sigma$ 

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  - One call to a SAT solver is enough to decide whether Σ defines y in terms of {x<sub>1</sub>,..., x<sub>n</sub>} (thanks to Padoa's theorem)

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- There is **no need to identify** f to compute  $\Sigma[y \leftarrow f(x_1, \dots, x_n)]$ 
  - The replacement phase can be replaced by an output variable elimination phase: if y ↔ f(x<sub>1</sub>,...,x<sub>n</sub>) is a gate of Σ, then

$$\Sigma[y \leftarrow f(x_1,\ldots,x_n)] \equiv \exists y.\Sigma$$

#### A two-step preprocessing

• "Detection = <u>B</u>ipartition": compute a **definability bipartition**  $\langle I, O \rangle$  of  $\Sigma$  such that  $I \cup O = Var(\Sigma), I \cap O = \emptyset$ , and  $\Sigma$  defines every variable  $o \in O$  in terms of I

#### A two-step preprocessing

- "Detection = <u>B</u>ipartition": compute a definability bipartition (*I*, *O*) of Σ such that *I* ∪ *O* = Var(Σ), *I* ∩ *O* = Ø, and Σ defines every variable *o* ∈ *O* in terms of *I*
- "Replacement =  $\underline{E}$ limination": compute  $\exists E.\Sigma$  for  $E \subseteq O$

#### A two-step preprocessing

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- "Replacement =  $\underline{E}$ limination": 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



$$u \leftrightarrow (x \land (y \lor z))$$

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

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

Identification:  $\Sigma \wedge \Sigma'_{\{x,y,z\}} \wedge u \wedge \overline{u'}$  is inconsistent  $\overline{X} \vee u \vee v$  $\overline{x} \vee \overline{y} \vee u$  $\overline{x} \vee \overline{z} \vee u$  $x \vee \overline{u}$  $v \lor z \lor \overline{u}$  $\overline{x} \lor u' \lor v'$  $\overline{x} \vee \overline{y} \vee u'$  $\overline{x} \vee \overline{z} \vee u'$  $x \vee \overline{u'}$  $v \lor z \lor \overline{u'}$  $\frac{u}{u'}$ 

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Identification: Flimination:  $\Sigma \wedge \Sigma'_{\{x,y,z\}} \wedge u \wedge \overline{u'}$  is inconsistent computing resolvents over *u*  $\overline{X} \vee u \vee v$ valid  $\overline{x} \lor v \lor x$  $\overline{x} \vee \overline{y} \vee u$  $\overline{x} \lor v \lor y \lor z$  $\overline{x} \vee \overline{z} \vee u$  $\overline{x} \vee \overline{y} \vee x$ valid  $x \vee \overline{u}$  $\overline{x} \vee \overline{y} \vee y \vee z$ valid  $v \lor z \lor \overline{u}$  $\overline{x} \lor \overline{z} \lor x$  valid  $\overline{x} \lor u' \lor v'$  $\overline{x} \lor \overline{z} \lor y \lor z$ valid  $\overline{x} \vee \overline{y} \vee u'$  $\overline{x} \vee \overline{z} \vee u'$  $x \vee \overline{u'}$  $v \lor z \lor \overline{u'}$ и

 $\frac{1}{u'}$ 

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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 (I, O) with |I| minimal
  - $\Rightarrow$  B is a greedy algorithm (one definability test per variable)
  - $\Rightarrow$  Only the minimality of *I* for  $\subseteq$  is guaranteed

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 (I, O) 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 Σ every variable of O but focusing on a subset E ⊆ O is enough
   ⇒ 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** 

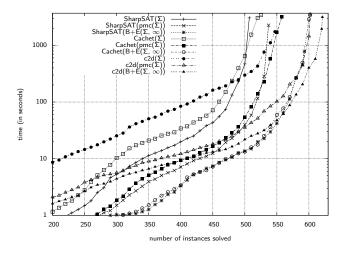
#### **Objectives:**

- 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

#### **Objectives:**

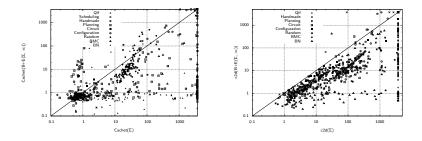
- Evaluating the computational benefits offered by B + E when used upstream to state-of-the-art model counters:
  - the search-based model counter Cachet
  - the search-based model counter SharpSAT
  - the compilation-based model counter C2D
- Comparing the benefits offered by B + E with those offered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

## **Empirical Results**



## **Empirical Results**

#### B + E vs. no preprocessing





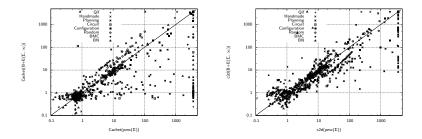
## FIGURE: Model counting time reductions achieved by $\mathbf{B} + \mathbf{E}$ vs. no preprocessing

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#### **Empirical Results**

B + E vs. pmc



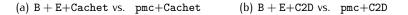


FIGURE: Model counting time reductions achieved by B + E vs. pmc

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- ► 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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