## On Definability for Model Counting

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- Key idea: Leveraging the power of modern SAT solvers to tackle other intractable problems
- ▶ Objective: Enlarging the sets of instances which can be solved in practice using "reasonable" resources
  - Knowledge compilers
  - MUS/MCS enumerators
  - OBF solvers
  - Model counters
- beyondnp.org



$$ightharpoonup \Sigma \mapsto \|\Sigma\| = ?$$



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$$\triangleright \ \Sigma = (x \vee y) \wedge (\neg y \vee z)$$



```
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```

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

```
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100
101
```

111



- $\Sigma \mapsto \|\Sigma\| = ?$
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- ▶ The models of  $\Sigma$  over  $\{x, y, z\}$  are :
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  - 111
- $||\Sigma|| = 4$



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



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On Definability for Model Counting

- ► 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):

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



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

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    - **.**..
  - Approximate model counters (SampleCount, etc.)
  - **...**
- ► In this talk: improving exact model counters by **preprocessing** the input

 $\mathtt{CNF} \to \mathtt{CNF}$ 



#### Preprocessings



- Objective: simplifying the input so that the task at hand can be achieved more efficiently from the input once preprocessed
- Simplifying = "reducing something"
- Trade-off preprocessing cost / rest of the computation to be looked for
- Using aggressive, computationally demanding preprocessing techniques can make sense when dealing with highly complex problems (like #SAT)
- P-preprocessing vs. NP-preprocessing





Similarities: two off-line approaches for improving the model counting task



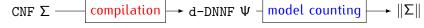


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  - ► "hard part" vs. "easy part"





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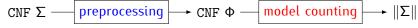






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preprocessing

► The two approaches can be **combined** 



## 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)

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- 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)
- **.**..



## Reducing What?



$$CNF \Sigma \mapsto CNF p(\Sigma)$$

- ▶ What are the connections between  $\Sigma$  and  $p(\Sigma)$ ?
- Removing clauses from Σ
- lacktriangle Removing literals in the clauses of  $\Sigma$
- **.**..



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

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- ▶ A clause  $\delta$  of a CNF  $\Sigma$  is redundant if and only if  $\Sigma \setminus \{\delta\} \models \delta$
- $\blacktriangleright$  A CNF  $\Sigma$  is irredundant if and only if it does not contain any redundant clause
- A subset  $\Sigma'$  of a CNF  $\Sigma$  is an irredundant equivalent subset (IES) of  $\Sigma$  if and only if  $\Sigma'$  is irredundant and  $\Sigma' \equiv \Sigma$
- lacktriangle Deciding whether a CNF  $\Sigma$  is irredundant is NP-complete
- ▶ Deciding whether a CNF  $\Sigma'$  is an irredundant equivalent subset (IES) of a CNF  $\Sigma$  is  $\mathsf{D}^p$ -complete
- Given an integer k, deciding whether a CNF  $\Sigma$  has an IES of size at most k is  $\Sigma_2^p$ -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



## Preserving What?



- ► Logical equivalence
- ► Queries over the input alphabet
- ► Number of models
- Satisfiability
- **.**..



# Measuring the Impact of a Preprocessing



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

- lacktriangle The number of variables in the input CNF  $\Sigma$
- ightharpoonup The size of Σ (the number of literals or the number of clauses in it)
- ightharpoonup The value of some 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  $\delta_2$ 

$$SE: (x_1 \vee x_2) \wedge (x_1 \vee x_2 \vee \bar{x}_3) \mapsto x_1 \vee x_2$$

- P-preprocessing
- Preserves logical equivalence
- Hence preserves the number of models of the input (over the original alphabet), its queries and its satisfiability
- $\blacktriangleright \# var(\Sigma) \geq \# var(\mathtt{SE}(\Sigma))$
- $\blacktriangleright \#lit(\Sigma) \geq \#lit(\mathtt{SE}(\Sigma))$





$$\Sigma = \begin{bmatrix} \overline{x} \lor u \lor v \\ \overline{x} \lor \overline{y} \lor u \\ \overline{x} \lor \overline{z} \lor u \\ x \lor \overline{u} \\ y \lor z \lor \overline{u} \end{bmatrix}$$



$$\overline{x} \lor u \lor v 
\overline{x} \lor \overline{y} \lor u$$

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

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



$$\Sigma = 
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\end{array}$$

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





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





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$$\overline{x} \lor u \lor v 
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$$\|\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
- Specific gates are typically sought for (literal equivalence, AND/OR gates, XOR gates) for complexity reasons
- ▶ 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 often used for efficiency reasons (P-preprocessing)



On Definability for Model Counting

#### <u> Literal Equivalence (LE)</u>



- Literal equivalence aims to detect equivalences between literals using BCP
- P-preprocessing
- $\blacktriangleright$  For each literal  $\ell$ , all the literals  $\ell'$  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  $\ell'$  occur in  $\Sigma$ , then those literals are found equivalent using BCP, but the converse does not hold)

#### Literal Equivalence (LE)

```
Algorithm 1: 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
3
          mark var(\ell);
          \mathcal{P}_{\ell} \leftarrow BCP(\Phi \cup {\{\ell\}});
          \mathcal{N}_{\ell} \leftarrow \text{BCP}(\Phi \cup \{\sim \ell\});
          \Gamma \leftarrow \{\ell \leftrightarrow \ell' | \ell' \neq \ell \text{ and } \ell' \in \mathcal{P}_{\ell} \text{ and } \sim \ell' \in \mathcal{N}_{\ell}\};
          // replacement
          foreach \ell \leftrightarrow \ell' \in \Gamma do
                 replace \ell by \ell' in \Phi;
```

return Φ



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# Literal Equivalence (LE): Example



$$\begin{split} \Sigma = & \\ & a \lor b \lor c \lor \neg d \quad \neg a \lor \neg b \lor \neg c \lor d \\ & a \lor b \lor \neg c \quad \neg a \lor \neg b \lor c \\ & \neg a \lor b \quad a \lor \neg b \\ & \neg e \lor \neg f \lor h \quad e \lor f \lor g \\ & e \lor \neg g \quad \neg e \lor \neg h \end{split}$$

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

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



#### Properties of LE



- Preserves the number of models (but not logical equivalence)
- $\blacktriangleright \# var(\Sigma) \geq \# var(LE(\Sigma))$
- $\blacktriangleright \#lit(\Sigma) \geq \#lit(LE(\Sigma))$



### LE: Reduction of the Number of Variables



equivSimpl( $\Sigma$ )

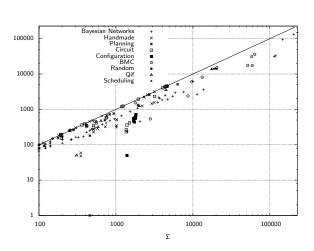


FIGURE – Comparing  $\#var(\Sigma)$  with  $\#var(LE(\Sigma))$ .



### LE: Reduction of the Size





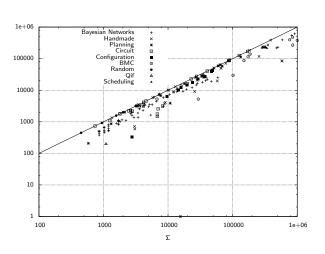


FIGURE – Comparing  $\#lit(\Sigma)$  with  $\#lit(LE(\Sigma))$ .



# Backbone Identification (BI)



- $\triangleright$  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  $\Sigma$  explicit, to conjoin it to  $\Sigma$ , and to use BCP (Boolean Constraint Propagation) on the resulting set of clauses
- NP-preprocessing



### Backbone Identification (BI)



### Algorithm 2: BI Backbone Identification

```
input: a CNF formula \Sigma output: the CNF 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} \text{ s.t. } \ell \notin \mathcal{B} \text{ do} 4 \mathcal{I}' \leftarrow \operatorname{solve}(\Sigma \cup \{ \sim \ell \}); 5 \mathsf{if} \ \mathcal{I}' = \emptyset \text{ then } \mathcal{B} \leftarrow \mathcal{B} \cup \{ \ell \} \text{else } \mathcal{I} \leftarrow \mathcal{I} \cap \mathcal{I}'; 6 return BCP(\Sigma \cup \mathcal{B})
```

# Backbone Identification (BI): Example



```
\Sigma =
       a \lor b
       \neg a \lor b
       \neg b \lor c
       c \lor d
       \neg c \lor e \lor f
       f \vee \neg g
```

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

$$BI(\Sigma) = b$$

$$c$$

$$e \lor f$$

$$f \lor \neg g$$



### Properties of BI



- Preserves logical equivalence
- $\#var(\Sigma) \ge \#var(BI(\Sigma))$
- ▶  $\#lit(\Sigma) \ge \#lit(BI(\Sigma))$



### BI: Reduction of the Number of Variables



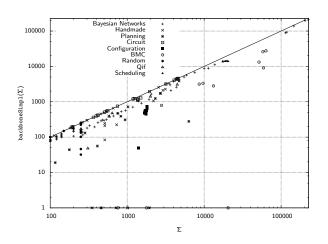


FIGURE – Comparing  $\#var(\Sigma)$  with  $\#var(BI(\Sigma))$ .



### BI: Reduction of the Size



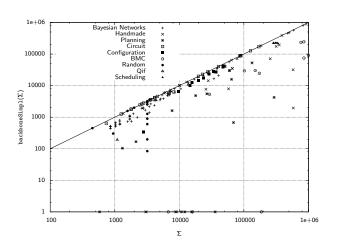


FIGURE – Comparing  $\#lit(\Sigma)$  with  $\#lit(BI(\Sigma))$ .



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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



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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

```
\begin{array}{ll} \text{literal equivalences} & y \leftrightarrow x_1 \\ \text{AND/OR gates} & y \leftrightarrow \left(x_1 \wedge \overline{x_2} \wedge x_3\right) \\ \text{XOR gates} & y \leftrightarrow \left(x_1 \oplus \overline{x_2}\right) \end{array}
```





- ► The (explicit) identification phase can be replaced by an implicit identification phase
- Stated otherwise, there is **no need to identify** f to determine that a gate of the form  $y \leftrightarrow f(x_1, \ldots, x_n)$  exists in  $\Sigma$





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













▶  $\Sigma$  **explicitly defines** y in terms of  $X = \{x_1, \dots, x_n\}$  iff there exists a formula  $f(x_1, \dots, x_n)$  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  $X = \{x_1, \dots, x_n\}$  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:  $\Sigma$  explicitly defines y in terms of X iff  $\Sigma$  implicitly defines y in terms of X

### Alessandro Padoa (1868-1937)





#### Padoa's theorem:

Let  $\Sigma_X'$  be equal to  $\Sigma$  where each variable but those of X have been renamed in a uniform way If  $y \not\in X$ , 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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  - Explicit definability = Implicit definability (Beth's theorem)



- ▶ There is **no need to identify** *f* to determine that a gate of the form  $y \leftrightarrow f(x_1, \dots, x_n)$  exists in  $\Sigma$ 
  - ► Gate identification = Explicit definability
  - Explicit definability = Implicit definability (Beth's theorem)
  - ightharpoonup One call to a SAT solver is enough to decide whether  $\Sigma$ defines y in terms of  $\{x_1, \ldots, x_n\}$  (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  $y \leftrightarrow f(x_1, ..., x_n)$  is a gate of  $\Sigma$ , then

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



# The B + E Preprocessing



### A two-step preprocessing

▶ "Identification =  $\underline{B}$ 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

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- ► "Replacement =  $\underline{E}$ limination": compute  $\exists E.\Sigma$  for  $E \subseteq O$



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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 (NP-preprocessing)

# Identifying u as an Output Variable and Eliminating it



#### Identification:

$$\Sigma \wedge \Sigma'_{\{x,y,z\}} \wedge u \wedge \overline{u'}$$
 is inconsistent

$$\begin{array}{c|c} \overline{x} \lor u \lor v \\ \overline{x} \lor \overline{y} \lor u \\ \overline{x} \lor \overline{z} \lor u \\ x \lor \overline{u} \\ y \lor z \lor \overline{u} \\ \overline{x} \lor u' \lor v' \\ \overline{x} \lor \overline{y} \lor u' \\ \overline{x} \lor \overline{z} \lor u' \\ x \lor \overline{u'} \\ y \lor z \lor \overline{u'} \\ \end{array}$$

# Identifying u as an Output Variable and Eliminating it



$$\Sigma \wedge \Sigma'_{\{x,y,z\}} \wedge u \wedge \overline{u'}$$
 is inconsistent

### Elimination:

computing resolvents over  $\boldsymbol{u}$ 

$\overline{x} \lor v \lor x$	valid
$\overline{x} \lor v \lor y \lor z$	
$\overline{x} \vee \overline{y} \vee x$	valid
$\overline{x} \vee \overline{y} \vee y \vee z$	valid
$\overline{X} \vee \overline{Z} \vee X$	valid
$\overline{X} \vee \overline{Z} \vee y \vee z$	valid



# Identifying u as an Output Variable and Eliminating it

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computing resolvents over u

$$\overline{x} \lor v \lor x \qquad \text{valid} 
\overline{x} \lor v \lor y \lor z 
\overline{x} \lor \overline{y} \lor x \qquad \text{valid} 
\overline{x} \lor \overline{y} \lor y \lor z \qquad \text{valid} 
\overline{x} \lor \overline{z} \lor x \qquad \text{valid} 
\overline{x} \lor \overline{z} \lor y \lor z \qquad \text{valid}$$

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



### 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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  - ⇒ B is a **greedy algorithm** (one definability test per variable)
  - $\Rightarrow$  Only the minimality of I for  $\subseteq$  is guaranteed
- $\triangleright$  It is not necessary to eliminate in  $\Sigma$  every variable of O but focusing on a subset  $E \subseteq O$  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 techniques (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 (used with -count -in\_memory -smooth\_all)
  - ▶ the compilation-based model counter D4



### Experiments



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  - ▶ the compilation-based model counter D4
- 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 Setting**



- 703 CNF instances from the SAT LIBrary
- 8 data sets: BN (Bayesian networks) (192), BMC (Bounded Model Checking) (18), Circuit (41), Configuration (35), Handmade (58), Planning (248), Random (104), 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



# Empirical Results: Reduction Achieved



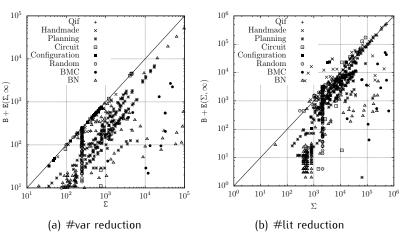
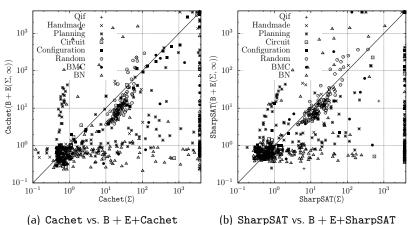


Figure – Reduction achieved by  $\mathtt{B} + \mathtt{E}$ 



# Empirical Results: Time Saving





(b) SharpSAT vs. B + E + SharpSAT

FIGURE – Time saved by using B + E upstream



# Empirical Results: Time Saving



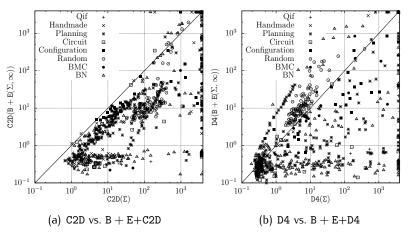
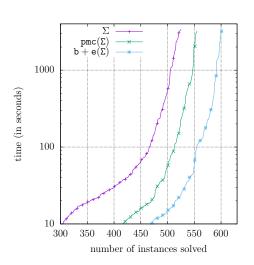


FIGURE – Time saved by using B+E upstream



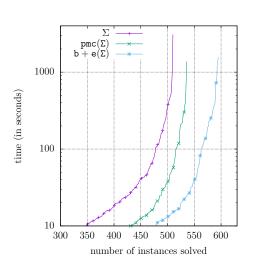




 ${\sf Figure}$  — Cachet depending on the preprocessing used







 $\label{eq:figure} \textbf{Figure} - \textbf{SharpSAT} \ depending \ on \ the \ preprocessing \ used$ 





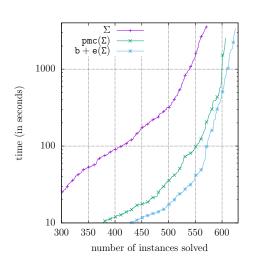


FIGURE - C2D depending on the preprocessing used





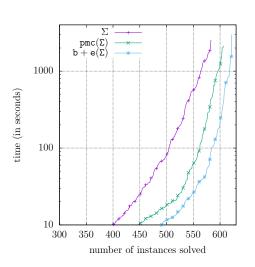


FIGURE - D4 depending on the preprocessing used



### Conclusion and Perspectives



#### Conclusion

- ▶ Design and implementation of the B + E preprocessor
- ▶ Empirical evaluation of B + E: for several model counters mc, mc(B + E(.)) proves computationally more efficient than mc(.)
- "Real" instances are structured ones



# Conclusion and Perspectives



#### Conclusion

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- "Real" instances are structured ones

### Perspectives

- Developing other ordering heuristics for B
- Investigating the connections to **projected model counting**: computing  $\|\exists E.\Sigma\|$  given a set E of variables and a formula  $\Sigma$

