

# SAT Preprocessing for MUS Extraction and MaxSAT

Anton Below

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# Abstract

State-of-the-art algorithms for industrial instances of MUS extraction problem and MaxSAT rely on iterative calls to a SAT solver. Preprocessing is crucial for the acceleration of SAT solving, and the key preprocessing techniques rely on the application of resolution and subsumption elimination. Additionally, satisfiability-preserving clause elimination procedures are often used. Since the computation typically involves a large number of SAT calls, an interesting question is whether an input instance to a problem can be preprocessed *up-front*, i.e. prior to running an MUS extractor or a MaxSAT solver, rather than (or, in addition to) during each iterative SAT solver call. The key requirement in this setting is that the preprocessing has to be *sound*, i.e. so that the solution to the original problem can be reconstructed correctly and efficiently after the execution of an algorithm on the preprocessed instance. In this talk we will examine some of the obstacles to such up-front preprocessing, and will discuss a solution that involves re-casting the MUS and MaxSAT computation problems in a so-called labelled-CNF framework.

# Outline

Introduction: MUSes, MaxSAT, SAT preprocessing

Direct reconstruction: techniques that work

Direct reconstruction: techniques that break

Sound preprocessing in Labelled CNF framework

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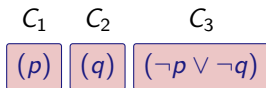
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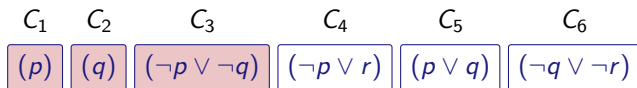
$M = \{C_1, C_2, C_3\}$  is UNSAT, and  $\forall C \in M, M \setminus \{C\}$  is SAT.

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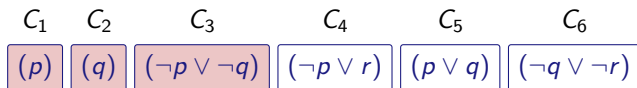


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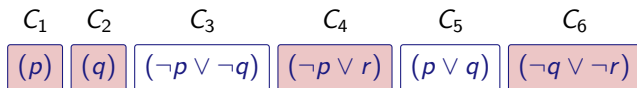


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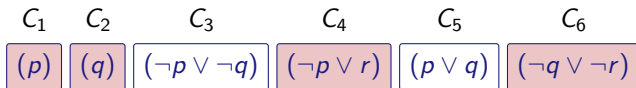


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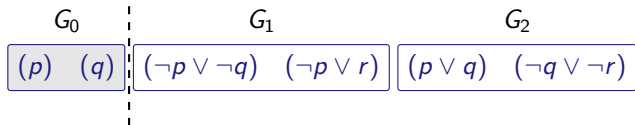
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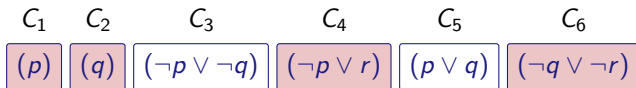
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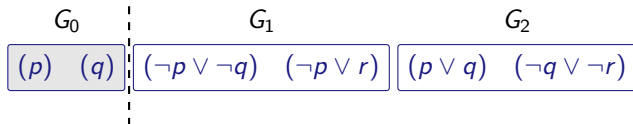
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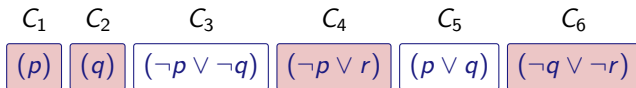
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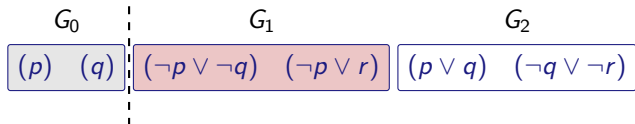
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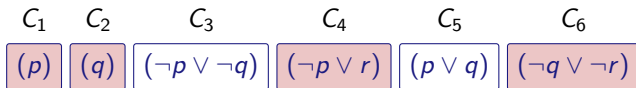


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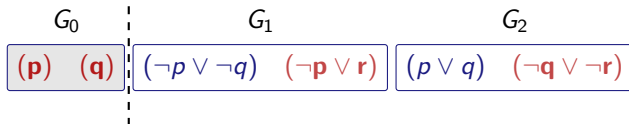
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A *MaxSAT solution* for  $F$ , is an assignment  $\tau$  that *maximizes*  $|Sat(F, \tau)|$ .  
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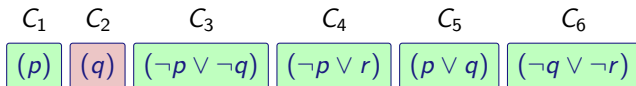
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## Weighted Partial MaxSAT:

- ▶ each clause  $C$  has a non-negative weight ( $\in \mathbb{N}^+$ ),  $w(C)$ ;
- ▶ some clauses are *hard* — *must* be satisfied (akin to group-0);
- ▶ weighted CNF (WCNF) formula:  $F = F^H \cup F^S$ ;
- ▶  $\tau$  is a weighted partial MaxSAT solution for WCNF  $F$  if  $\tau(F^H) = 1$ , and  $\tau$  minimizes  $\text{cost}(\tau) = \sum_{C \in \text{Unsat}(F^S, \tau)} w(C)$ .

## Applications

- ▶ Identification and repair of sources of inconsistency
  - circuit error diagnosis; error localization in product configuration.
- ▶ Identification of relevant features of systems:
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## Computation of (group-)MUSes and MaxSAT solutions

- ▶ Algorithms aimed at industrial instances are based on iterative calls to a SAT solver.
- ▶ SAT solving is the main bottleneck.
- ▶ Number of SAT calls is a function of the size of the input formula.
- ▶ In the SAT world: *preprocessing* is essential for efficient SAT solving.

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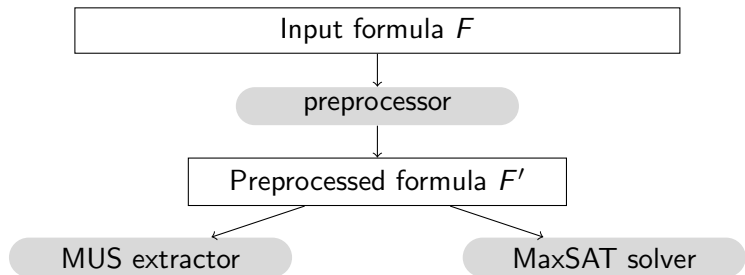
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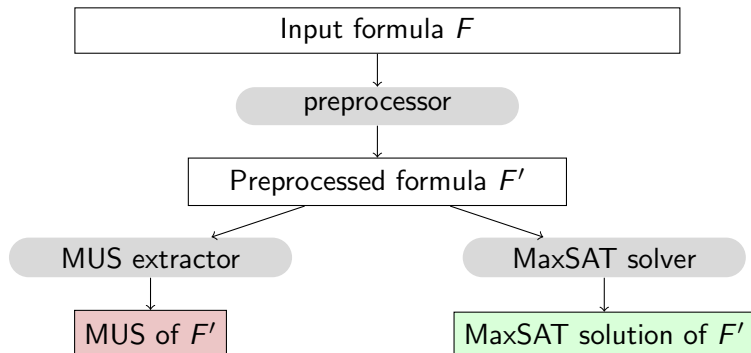
replace  $(x \vee C) (\neg x \vee C \vee D)$  with  $(x \vee C) (C \vee D)$ .

Important: a model of original formula can be reconstructed *efficiently*.

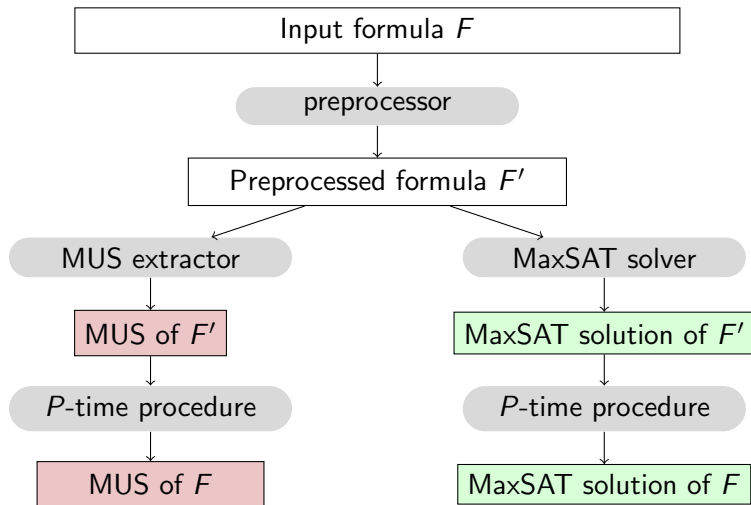
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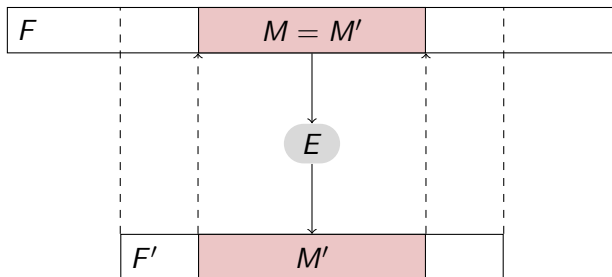
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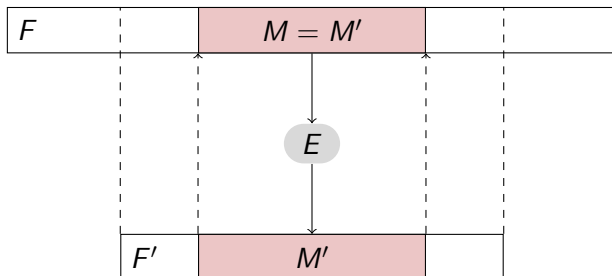
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# Plain MUS: clause elimination



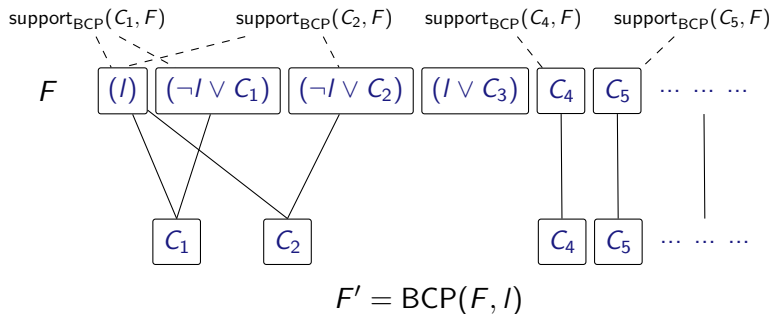
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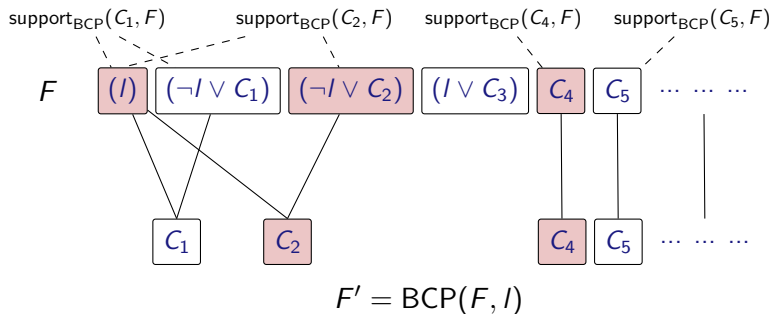
Prop: Any MUS of  $F' = E(F)$  is an MUS of  $F$ .

Reconstruction is trivial.

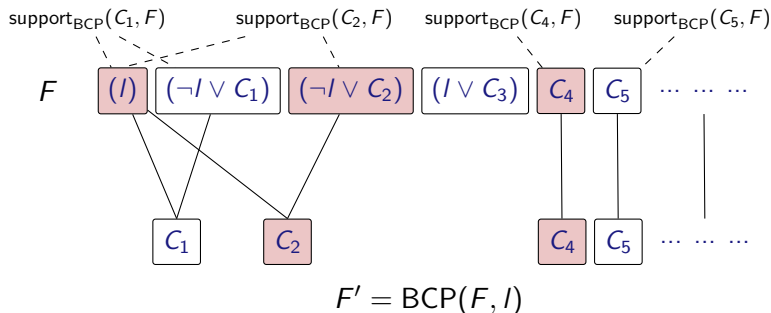
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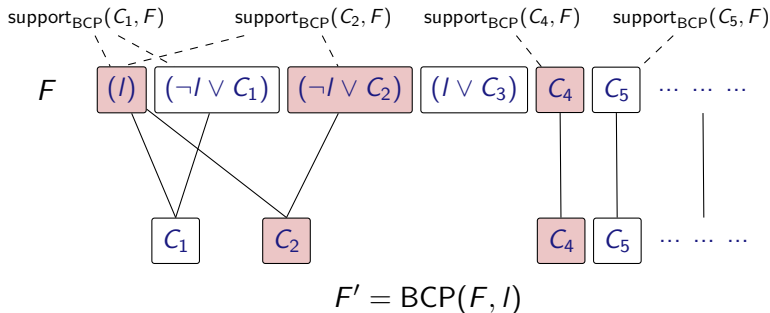


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Prop: If  $M'$  is an MUS of  $F' = \text{BCP}(F, I)$ , then  
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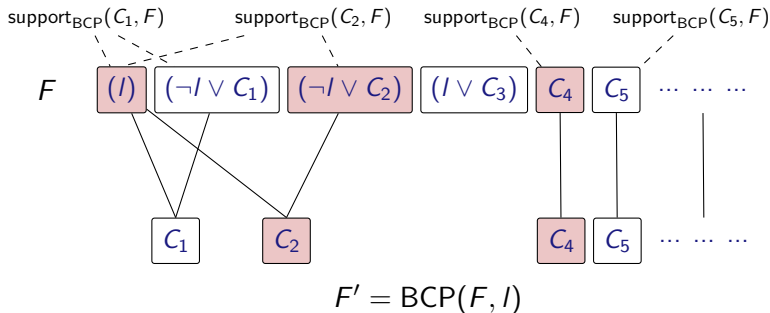


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Pf: Take a witness  $\tau$  for  $C_2$  in  $F'$ . Then,  $\tau \cup \{I\}$  is a witness for  $(\neg I \vee C_2)$ , while  $\tau \cup \{\neg I\}$  is a witness for  $(I)$ . If  $\tau$  is a witness for  $C_4$  in  $F'$ , then  $\tau \cup \{I\}$  is a witness for  $C_4$  in  $F$ .

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Reconstruction can be done in P-time.

## Group-MUS: *monotone* clause elimination

Def: A clause elimination procedure  $E$  is *monotone* iff for any  $F' \subseteq F$ ,  $E(F') \subseteq E(F)$ .

Example: BCE — if  $C$  is blocked in  $F$ , its blocked in any  $F' \subseteq F$ .

Non-example: SUB — if  $C_1 \subset C_2$  in  $F$ , but  $C_1$  is not in  $F'$ .



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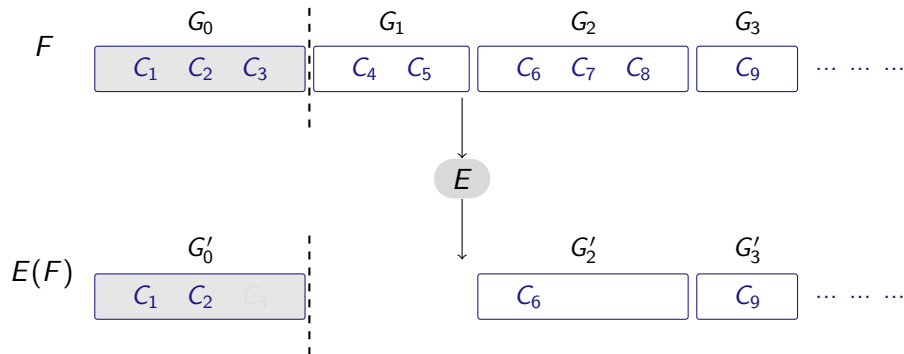
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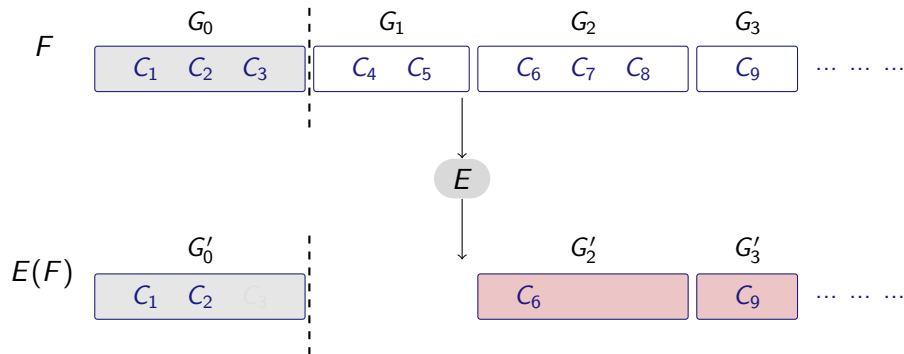
Prop: If  $E$  is monotone, then  $\text{MUS}(E(F)) = \text{MUS}(F)$ , i.e. any monotone clause elimination procedure is *MUS-preserving*.

Pf: if  $M \in \text{MUS}(E(F))$ , then  $M \subseteq E(F) \subseteq F$ , i.e.  $M \in \text{MUS}(F)$  (doesn't matter that  $E$  is monotone); if  $M \in \text{MUS}(F)$ , then  $E(M) = M$  because  $E$  is SAT-preserving, and since  $M \subseteq F$ , we have  $E(M) \subseteq E(F)$  by monotonicity, i.e.  $M \in \text{MUS}(E(F))$ .

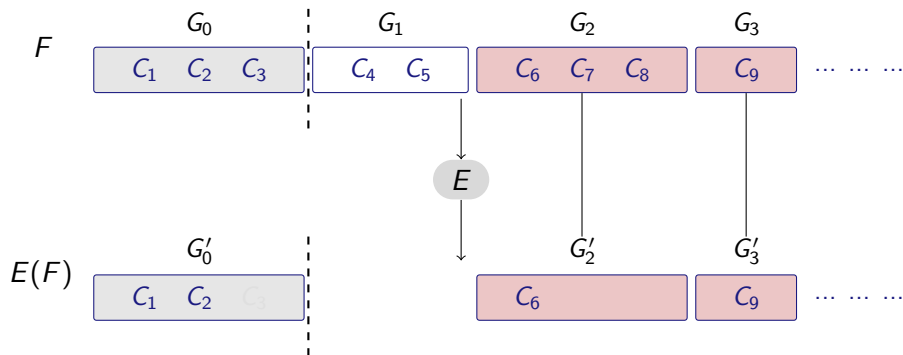
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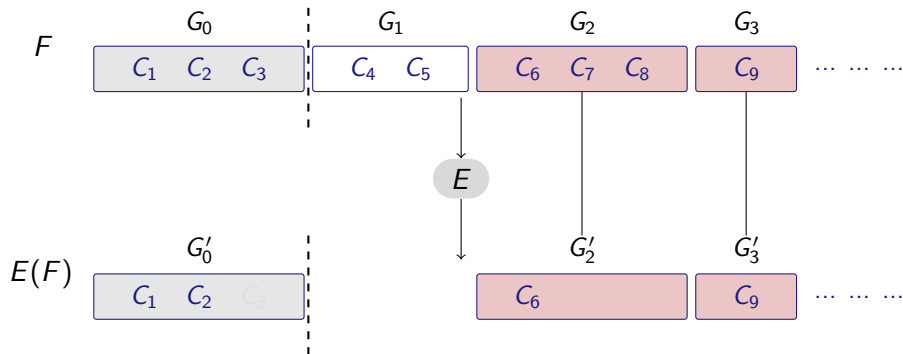


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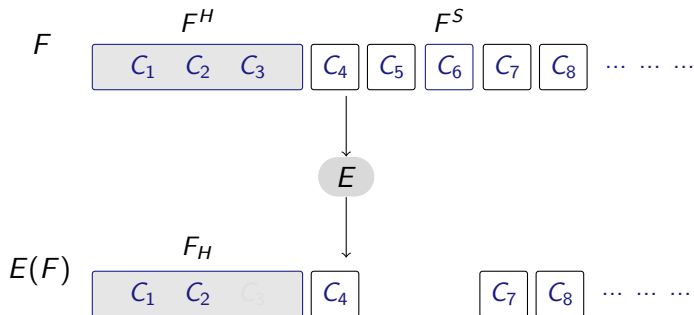


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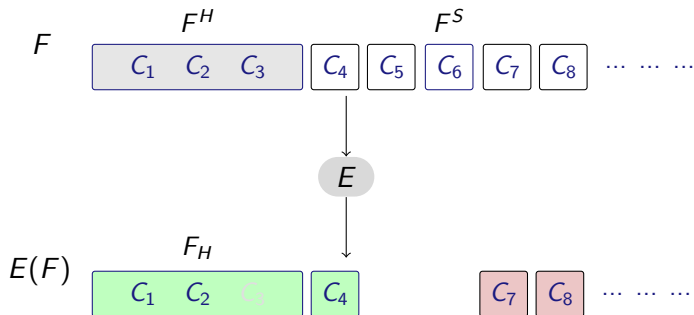
Prop: If  $E$  is MUS-preserving, then  $M$  is a group-MUS of  $F$ .

Note: in particular, this is true if  $E$  is monotone (e.g. BCE).

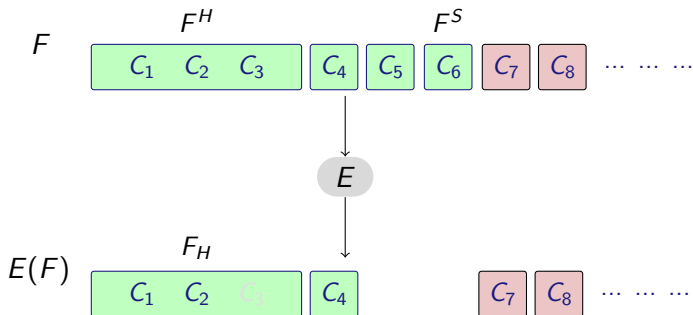
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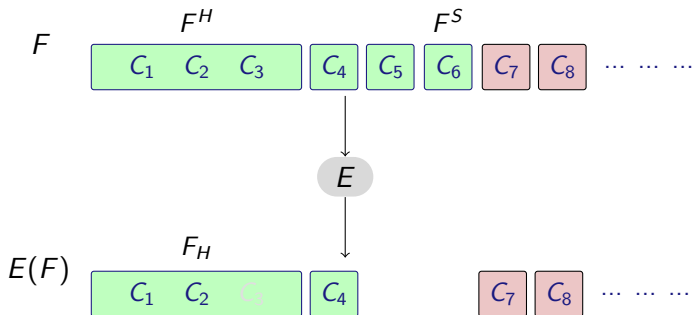


Prop: If  $E$  is MUS-preserving, and  $\tau$  is a MaxSAT solution for  $E(F)$ , then  $\alpha_E(\tau)$  is a MaxSAT solution for  $F$ .

Pf:  $E$  is MUS-preserving  $\Rightarrow E$  is MCS-preserving (HS-duality), and so min-cost MCS is the same. Only MSS clauses will be eliminated.  $\tau$  is a model of the MSS of  $E(F)$ ,  $\alpha_E(\tau)$  must be a model of corresponding MSS of  $F$ .



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**Note:** in particular, this is true if  $E$  is monotone (e.g. BCE).

# Outline

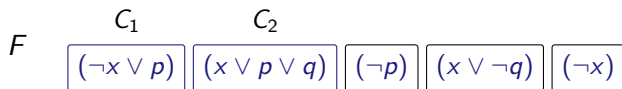
Introduction: MUSes, MaxSAT, SAT preprocessing

Direct reconstruction: techniques that work

Direct reconstruction: techniques that break

Sound preprocessing in Labelled CNF framework

# Plain MUS: SSR

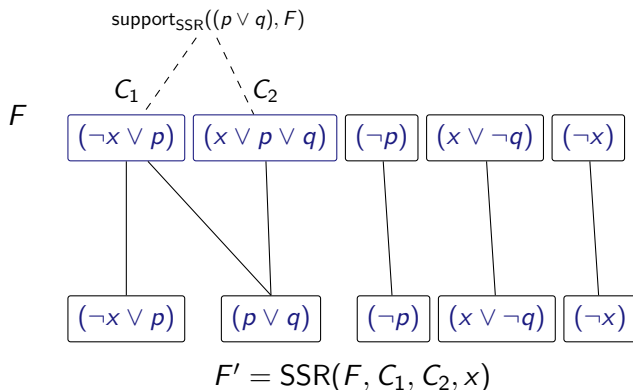


$$F \quad \begin{array}{cc} C_1 & C_2 \\ \boxed{(\neg x \vee p)} & \boxed{(x \vee p \vee q)} & \boxed{(\neg p)} & \boxed{(x \vee \neg q)} & \boxed{(\neg x)} \end{array}$$

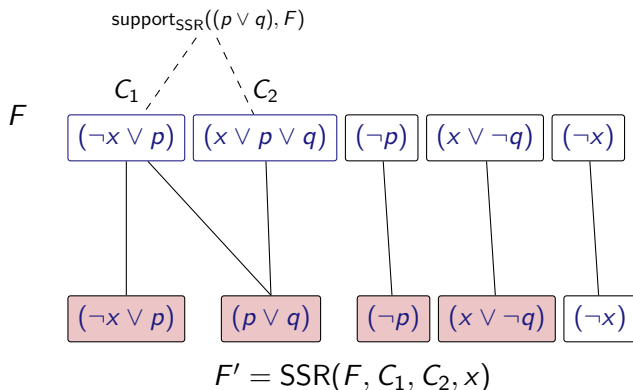
$$\boxed{(\neg x \vee p)} \quad \boxed{(p \vee q)} \quad \boxed{(\neg p)} \quad \boxed{(x \vee \neg q)} \quad \boxed{(\neg x)}$$

$$F' = \text{SSR}(F, C_1, C_2, x)$$

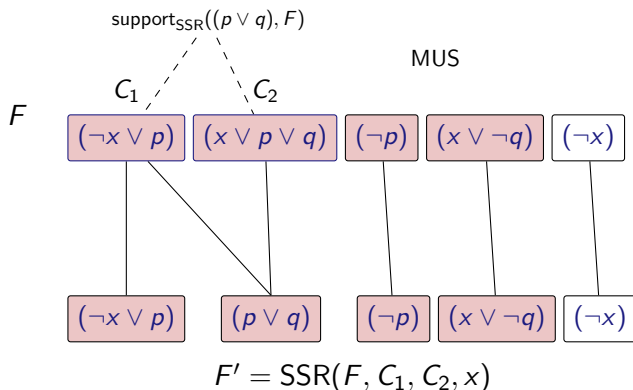
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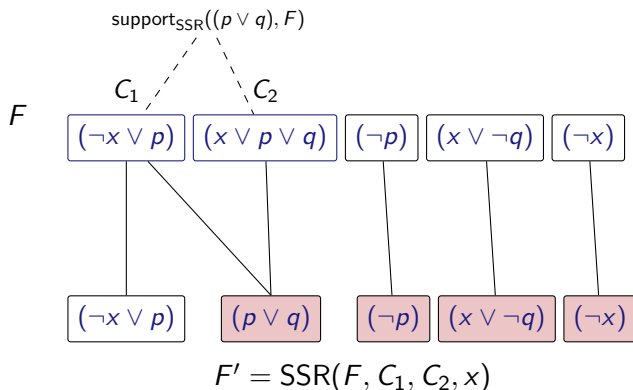
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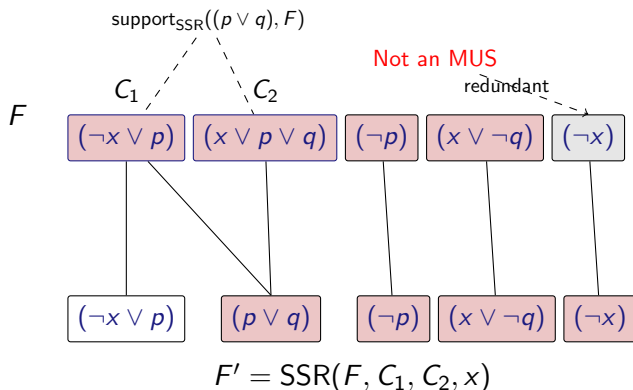


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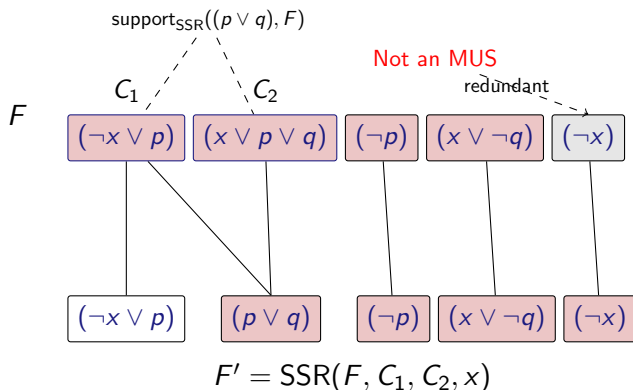




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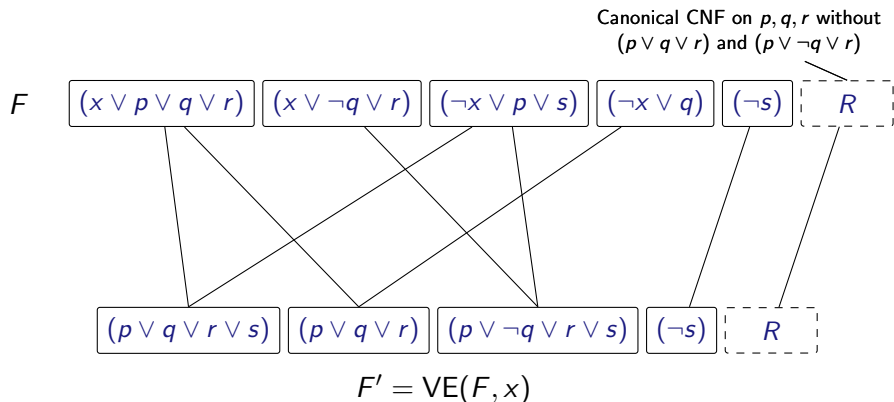
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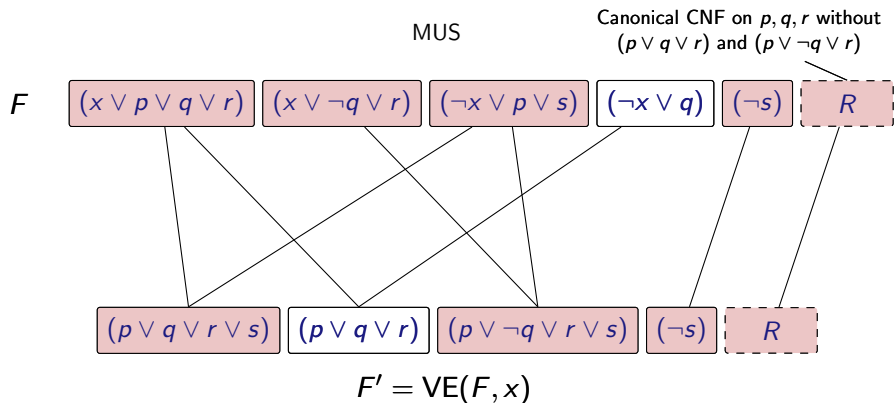
For an MUS  $M'$  of  $F' = \text{SSR}(F, C, D, I)$ ,

$M = \bigcup_{C \in M'} \text{support}_{\text{SSR}}(C, F)$  might be **not** an MUS of  $F$ .

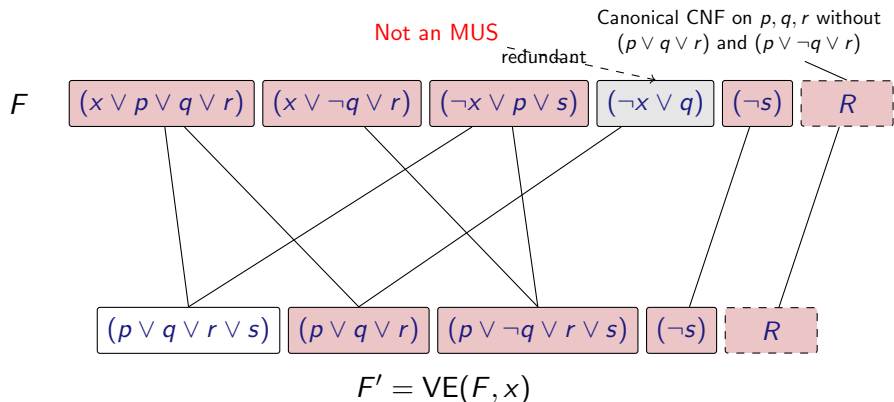
# Plain MUS: VE



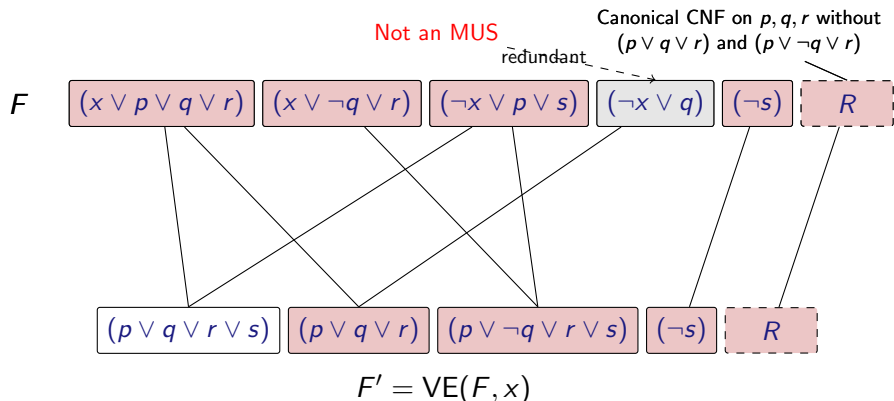
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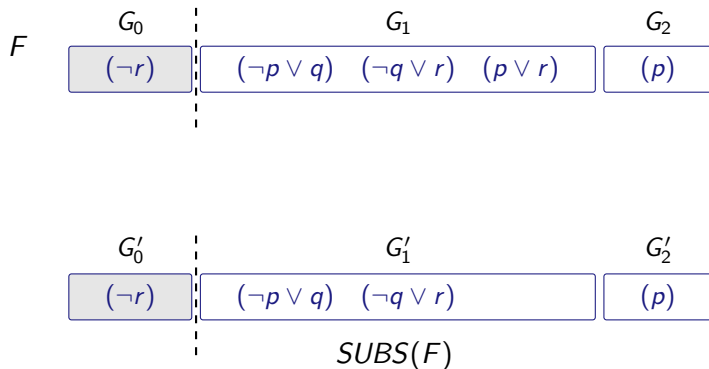


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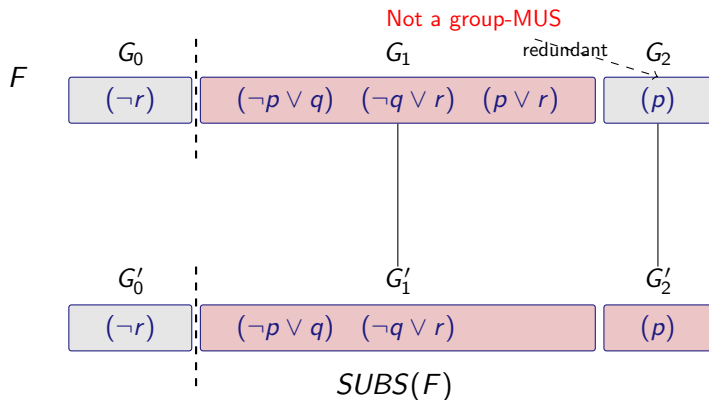
$M = \bigcup_{C \in M'} \text{support}_{\text{VE}}(C, F)$  might be **not** an MUS of  $F$ .

Doesn't work even if  $M$  is greedily minimized to avoid duplicate resolvents.

# Group-MUS: subsumption

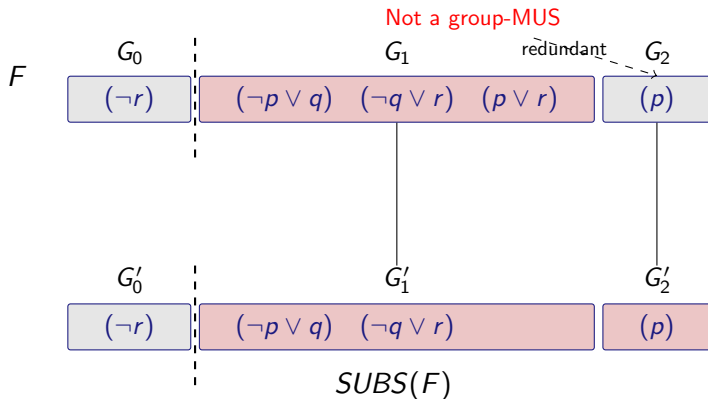


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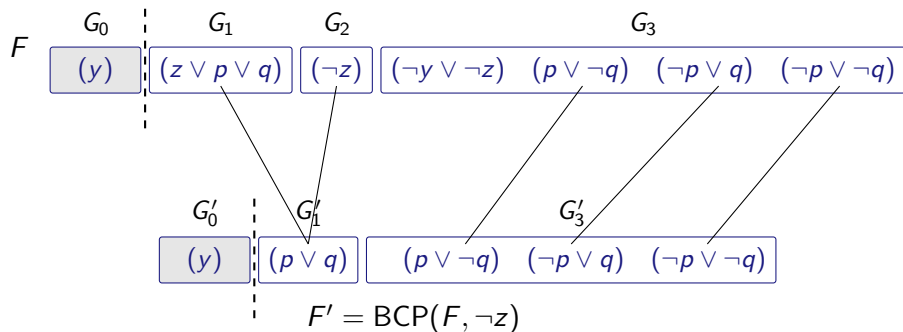


For a group-MUS  $M'$  of  $F' = E(F)$ ,

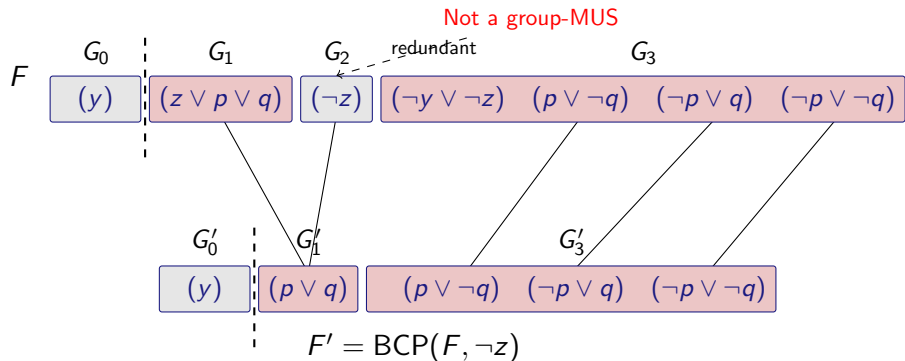
$M = \{G_i \in F \mid G'_i \in F'\}$  might be **not** a group-MUS of  $F$ .

Note: can do *certain* things with subsumption: eliminate *within* a group, or from group-0 to other groups.

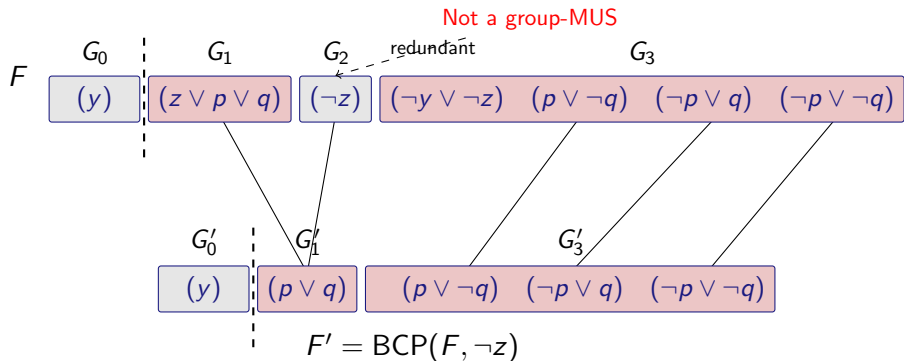
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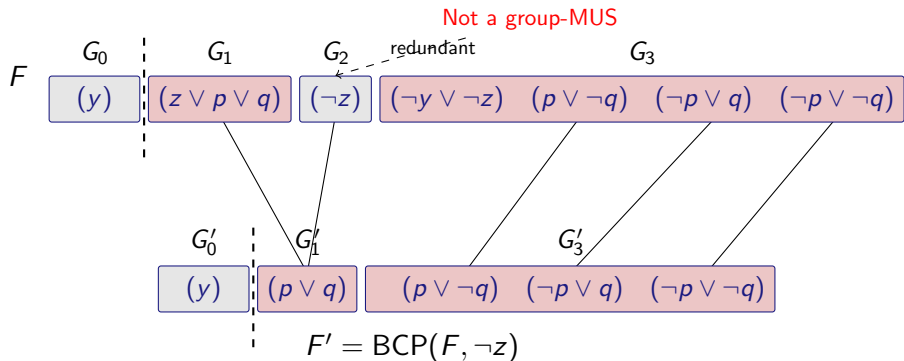
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BCP in the group-MUS setting is problematic.

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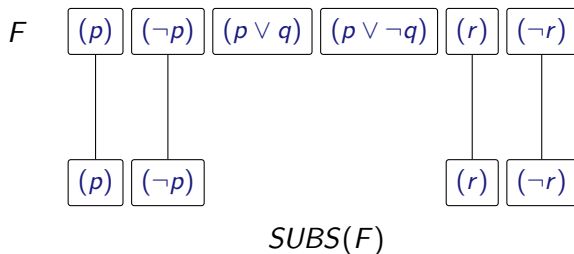
Note: again, can do *certain* things with BCP: propagate *within* a group, or from group-0 to other groups.

Note: VE and SSR didn't work for plain MUSes, and so won't work for group-MUSes either.

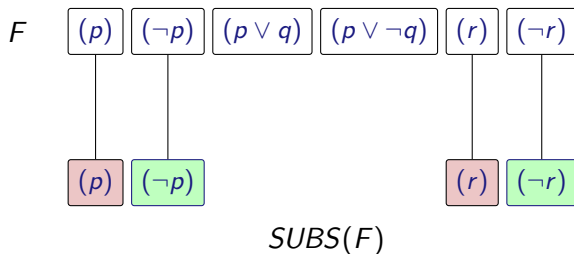
# MaxSAT: subsumption

$$F \quad \boxed{(p)} \quad \boxed{(\neg p)} \quad \boxed{(p \vee q)} \quad \boxed{(p \vee \neg q)} \quad \boxed{(r)} \quad \boxed{(\neg r)}$$

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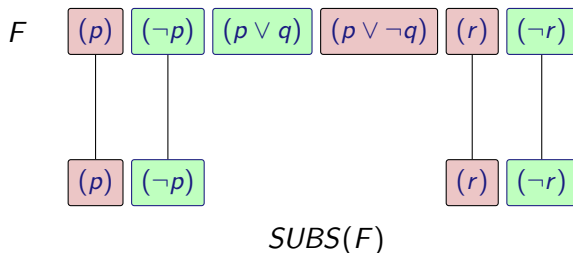
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$\tau = \{\neg p, \neg r\}$  is a MaxSAT solution for  $SUBS(F)$ ,  $cost(\tau) = 2$ .



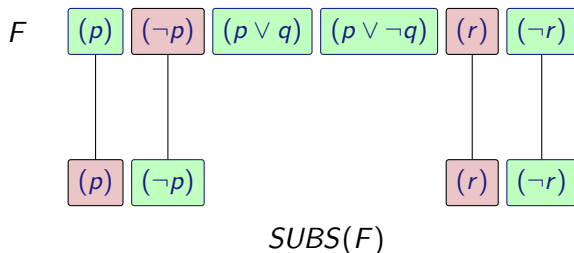
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$\tau' = \{\neg p, \neg r, q\}$  is not a MaxSAT solution for  $F$ ,  $cost(\tau') = 3$ , but  $cost(\{p, \neg r, q\}) = 2$ .

# MaxSAT: subsumption



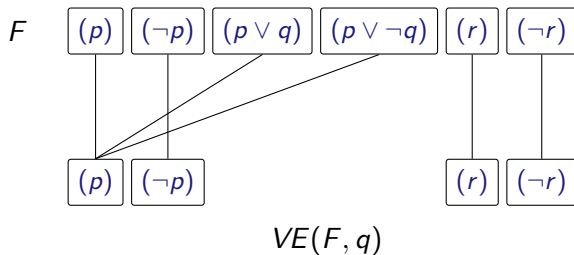
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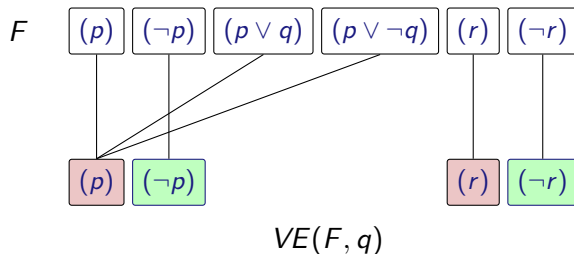
Reason: subsumption may remove some of the MUSes of  $F$ .

$$F \quad (p) \quad (\neg p) \quad (p \vee q) \quad (p \vee \neg q) \quad (r) \quad (\neg r)$$

# MaxSAT: VE

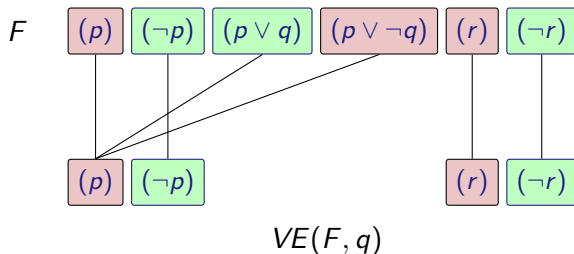


# MaxSAT: VE



$\tau = \{\neg p, \neg r\}$  is a MaxSAT solution for  $VE(F, q)$ ,  $cost(\tau) = 2$ .

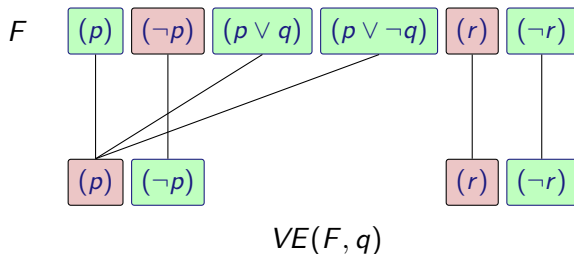
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Reason: VE *changes* the MUSes of  $F$ . Additional problem: what to do with weights ?

# MaxSAT: resolution-based preprocessing

None of the resolution-based techniques are *sound* for MaxSAT, because the resolution rule itself is not sound.

$$(x \vee C_1) \otimes_x (\neg x \vee C_2) = (C_1 \vee C_2)$$

Consider an assignment  $\tau$  s.t.  $\tau(\neg x \vee C_2) = 0$  and  $\tau(C_1) = 1$ .



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MaxSAT Resolution rule [Bonet et al, 07]

$$\begin{array}{l} x \vee a_1 \vee \dots \vee a_s \\ \bar{x} \vee b_1 \vee \dots \vee b_t \\ \hline a_1 \vee \dots \vee a_s \vee b_1 \vee \dots \vee b_t \\ x \vee a_1 \vee \dots \vee a_s \vee \bar{b}_1 \\ x \vee a_1 \vee \dots \vee a_s \vee b_1 \vee \bar{b}_2 \\ \dots \\ x \vee a_1 \vee \dots \vee a_s \vee b_1 \vee \dots \vee b_{t-1} \vee \bar{b}_t \\ \bar{x} \vee b_1 \vee \dots \vee b_t \vee \bar{a}_1 \\ \bar{x} \vee b_1 \vee \dots \vee b_t \vee a_1 \vee \bar{a}_2 \\ \dots \\ \bar{x} \vee b_1 \vee \dots \vee b_t \vee a_1 \vee \dots \vee a_{s-1} \vee \bar{a}_s \end{array}$$

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Although sound and complete for MaxSAT solving, is not suitable for preprocessing.

# Summary of Direct Reconstruction

## Techniques that work

- ▶ plain-MUS: any clause elimination, BCP.
- ▶ group-MUS: MUS-preserving (sp. monotone) clause elimination.
- ▶ MaxSAT: MUS-preserving (sp. monotone) clause elimination.

## Techniques that break

- ▶ plain-MUS: VE, SSR
- ▶ group-MUS: SUB, BCP, VE, SSR
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Note: Some special cases do work:

- ▶ group-MUS: within groups, or from group-0 to other groups.
- ▶ MaxSAT: within  $F^H$ , or from  $F^H$  to soft clauses.

Need a generic framework for guaranteed correctness-preserving application of preprocessing techniques.

# Outline

Introduction: MUSes, MaxSAT, SAT preprocessing

Direct reconstruction: techniques that work

Direct reconstruction: techniques that break

Sound preprocessing in Labelled CNF framework

# Labelled CNF (LCNF) formulas

Original motivation: generalize group-MUS to *intersecting* groups

Note: those arise naturally in circuit-MUS and variable-MUS settings.

Note: generalizes other related problems, such as MCS computation and MaxSAT.

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## The components of the framework

- ▶  $LbIs$ : a countable set of *labels*.
- ▶ *Labelled clause*  $C^L$  is a tuple  $\langle C, L \rangle$ , where  $C$  is a clause, and  $L \subset LbIs$  is a finite set of labels.
- ▶ *LCNF formula*  $\Phi$  is a set of labelled clauses.
  - $Cls(\Phi)$  - the “normal” clauses of  $\Phi$ , i.e.  $\bigcup_{C^L \in \Phi} C$ .
  - $LbIs(\Phi)$  - the labels of  $\Phi$ , i.e.  $\bigcup_{C^L \in \Phi} L$ .



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  - $Lbls(\Phi)$  - the labels of  $\Phi$ , i.e.  $\bigcup_{C^L \in \Phi} L$ .

Example ( $Lbls = \mathbb{N}$ ):

$$\Phi \quad \boxed{(x)^\emptyset} \quad \boxed{(y)^\emptyset} \quad \boxed{(z \vee p \vee q)^{\{1\}}} \quad \boxed{(p \vee \neg q)^{\{2\}}} \quad \boxed{(\neg p \vee q)^{\{2,3\}}} \quad \boxed{(\neg p \vee \neg q)^{\{3\}}}$$

## LCNF formulas: induced subformulas

Def: For an LCNF  $\Phi$ , let  $M \subseteq \text{Lbls}(\Phi)$ . Then,  $\Phi|_M = \{C^L \in \Phi \mid L \subseteq M\}$  is a subformula of  $\Phi$  *induced* by  $M$ .

Alternatively: any clause that has a label *outside* of  $M$  is removed from  $\Phi$ .

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**Note:** Clauses with the label-set  $\emptyset$  cannot be removed  $\Rightarrow$  a convenient way to represent  $G_0$  clauses in group-MUS, or  $F^H$  in MaxSAT.

# LCNF formulas: minimal unsatisfiability and MUSes

Satisfiability: LCNF  $\Phi$  is SAT iff  $Cls(\Phi)$  is SAT.

Def:  $M \subseteq Lbls(\Phi)$  is an *minimal unsatisfiable subformula (MUS)* of  $\Phi$  if  $\Phi|_M \in \text{UNSAT}$ , and  $\forall l \in M, \Phi|_{M \setminus \{l\}} \in \text{SAT}$

I.e. the removal of *any* label from  $\Phi|_M$  makes it SAT.

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Natural mapping between CNF/group-CNF and LCNF:

$C_1$	$C_2$	$C_3$	$C_4$	$C_5$	$C_6$
$(p)$	$(q)$	$(\neg p \vee \neg q)$	$(\neg p \vee r)$	$(p \vee q)$	$(\neg q \vee \neg r)$

MUS is  $\{C_1, C_2, C_4, C_6\}$

$\Phi$	$(p)^{\{1\}}$	$(q)^{\{2\}}$	$(\neg p \vee \neg q)^{\{3\}}$	$(\neg p \vee r)^{\{4\}}$	$(p \vee q)^{\{5\}}$	$(\neg q \vee \neg r)^{\{6\}}$
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LMUS is  $\{1, 2, 4, 6\}$

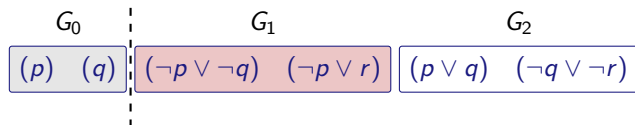
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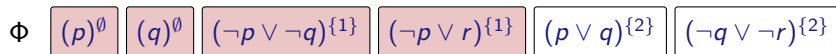
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Natural mapping between CNF/group-CNF and LCNF:



group-MUS is  $\{G_1\}$



LMUS is  $\{1\}$

# LCNF formulas: MCSes and MaxSAT

Def:  $S \subseteq Lbls(\Phi)$  is a *maximal satisfiable subformula (MSS)* of  $\Phi$  if  $\Phi|_S \in \text{SAT}$ , and  $\forall I \in Lbls(\Phi) \setminus S, \Phi|_{S \cup \{I\}} \in \text{UNSAT}$ .

Def:  $R \subseteq Lbls(\Phi)$  is a *minimal correction subset (MCS)* of  $\Phi$  if  $\Phi|_{Lbls(\Phi) \setminus R} \in \text{SAT}$ , and  $\forall I \in R, \Phi|_{(Lbls(\Phi) \setminus R) \cup \{I\}} \in \text{UNSAT}$ .

Note: hitting sets duality between MUSes and MCSes holds for LCNFs.



# LCNF formulas: MCSes and MaxSAT

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Def:  $R \subseteq Lbls(\Phi)$  is a *minimal correction subset (MCS)* of  $\Phi$  if  $\Phi|_{Lbls(\Phi) \setminus R} \in \text{SAT}$ , and  $\forall I \in R, \Phi|_{(Lbls(\Phi) \setminus R) \cup \{I\}} \in \text{UNSAT}$ .

Note: hitting sets duality between MUSes and MCSes holds for LCNFs.

MaxSAT for LCNFs:

- ▶ each *label*  $I$  has a *weight*  $w(I) \in \mathbb{N}^+$ ;
- ▶ the *cost* of a set of labels  $L$  is  $\sum_{I \in L} w(I)$ .
- ▶ weighted LCNF: just an LCNF + the weights for labels.
- ▶ a *MaxSAT solution* for  $\Phi$  is a model of an MSS of the *maximum cost*. Alternatively: falsifies only the clauses of an MCS of the *minimum cost*.

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Note: a natural mapping between MaxSAT instances and weighted LCNFs.

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## Subsumption elimination

Theorem: Let  $C_1^{L_1}$  and  $C_2^{L_2}$  in  $\Phi$  be such that  $C_1 \subseteq C_2$  and  $L_1 \subseteq L_2$ . Then, any MUS of  $\Phi \setminus \{C_2^{L_2}\}$  is an MUS of  $\Phi$ , and vice versa.

Note: by hitting sets duality,  $\text{MCS}(\Phi \setminus \{C_2^{L_2}\}) = \text{MCS}(\Phi)$ . Any MaxSAT solution  $\tau$  for  $\Phi \setminus \{C_2^{L_2}\}$  is a MaxSAT solution for  $\Phi$ .

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- ▶ I.e. subsumption is OK, as long as the labels also have a subset relationship.
- ▶ Effectively, blocks subsumption across groups in group-MUS, and across soft clauses in MaxSAT (good).
- ▶ Blocks subsumption in plain MUS instances: a weaker condition that allows for it seems to be possible.

## LCNF formulas: preprocessing

Def: Resolution:  $(x \vee C_1)^{L_1} \otimes_x (\neg x \vee C_2)^{L_2} = (C_1 \vee C_2)^{L_1 \cup L_2}$ .

Def: Variable elimination:  $VE(\Phi, x) = \Phi \cup (\Phi_x \otimes_x \Phi_{\neg x}) \setminus (\Phi_x \cup \Phi_{\neg x})$ .

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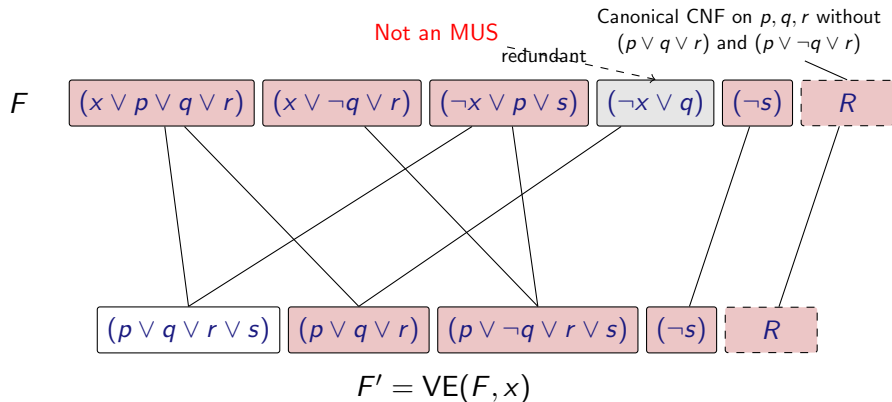
Note: by hitting sets duality,  $MCS(VE(\Phi, x)) = MCS(\Phi)$ .

For any MaxSAT solution  $\tau$  for  $VE(\Phi, x)$  there is a MaxSAT solution for  $\Phi$  with exactly the same cost (exactly the same MCS).

To get a solution for  $\Phi$  (the model of the MSS) do the standard reconstruction after VE.



# Remember the bad example for VE ?



For an MUS  $M'$  of  $F' = VE(F, x)$ ,

$M = \bigcup_{C \in M'} \text{support}_{VE}(C, F)$  might be **not** an MUS of  $F$ .

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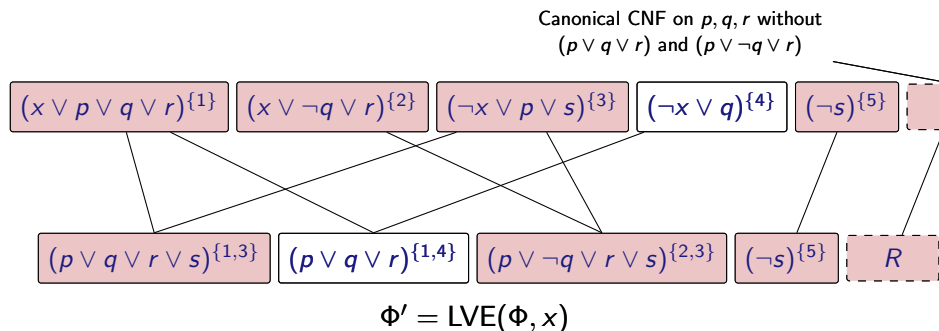
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Theorem: Any LMUS of  $LVE(\Phi, x)$  is an LMUS of  $\Phi$ , and vice versa.



Label 4 is redundant ! The LMUS of  $\Phi'$  is  $\{1, 2, 3, 5, \dots\}$  – same as for  $\Phi$ .

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BCP and SSR can also be defined for LCNFs so that the correctness is preserved.

Bottom line: labels enable correctness-preserving preprocessing for MUS and group-MUS extraction, and for computing MaxSAT solutions.

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## Preprocessing flow

Given  $F$ , a (group-)MUS instance or an instance of MaxSAT:

1. Run any MUS-preserving clause elimination procedure (like BCE).
2. For plain-MUSes may also run SUB and BCP (while keeping trace).
3. Convert the resulting formula  $F_1$  to LCNF  $\Phi_1$ .
4. Run VE, SSR on  $\Phi_1$  to get  $\Phi_2$ .
5. Compute an MUS  $M$  of  $\Phi_2$  or a MaxSAT solution  $\tau$  for  $\Phi_2$ .
6. The labels in  $M$  represent a (group-)MUS of  $F_1$ . The cost of a MaxSAT solution of  $F_1$  is the same as the cost of  $\tau$ ; to get a solution do the standard reconstruction after VE.

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## LCNF formulas: computing MUSes

Given an LCNF  $\Phi$ , for each label  $l_i \in Lbls(\Phi)$ ,  $1 \leq i \leq n$ , introduce a fresh variable  $p_i$ .

Then, convert each labelled clause  $C^{\{l_1, \dots, l_k\}}$  in  $\Phi$  into a clause  $(p_1 \vee \dots \vee p_k \vee C)$ . Let  $F_\Phi$  denote the resulting CNF.

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### Reduction to group-MUS problem

Let  $G_0 = F_\Phi$ , and for each  $l_i \in Lbls(\Phi)$ , let  $G_i = \{(\neg p_i)\}$ .

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## Direct computation

Load  $F_\Phi$  into an incremental SAT solver, and use  $p_i$ 's as assumptions.

Note: this is how most of the MUS extractors compute MUSes and group-MUSes anyway.

## LCNF formulas: computing MaxSAT solutions

Given a weighted LCNF  $\Phi$ , for each label  $l_i \in Lbls(\Phi)$ ,  $1 \leq i \leq n$ , introduce a fresh variable  $p_i$ .

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Let  $F^H = F_\Phi$ , and for each  $l_i \in Lbls(\Phi)$ , create a soft clause  $C_i = (\neg p_i)$  with weight  $w(l_i)$ .

Compute a MaxSAT solution  $\tau$  of  $F^H \cup C_1 \cup \dots \cup C_n$ .  $\tau$  is a MaxSAT solution of  $\Phi$ .

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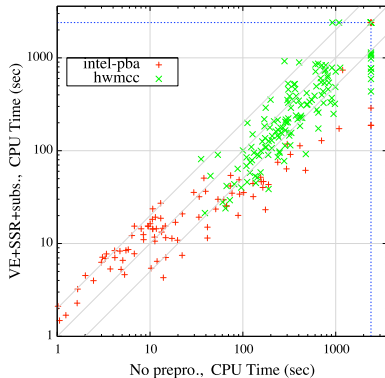
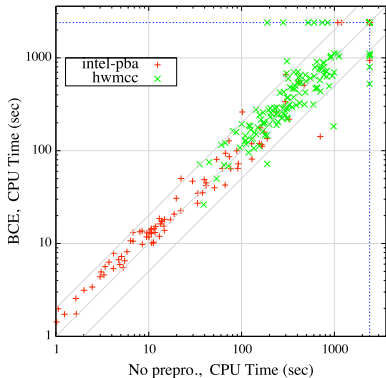
## Direct computation

Load  $F_\Phi$  into an incremental SAT solver, and use  $p_i$ 's as assumptions, and run a modified MaxSAT algorithm (need to “relax”  $p_i$ 's, not clauses).

Note: this also gives a nice incremental-SAT based MaxSAT algorithm.

# Experimental evaluation: group-MUS

- ▶ 99 group-CNF instances from Intel (PBA), 148 instances derived from HWMCC.
- ▶ Time limit 1800 sec, memory limit 4 GB.



# Experimental evaluation: MaxSAT

- ▶ 1000 instances from MaxSAT Competition 2013.
- ▶ Time limit 1800 sec, memory limit 4 GB.

	MaxSAT		Partial MaxSAT		Weighted Partial MaxSAT	
	#Sol.	A.CPU	#Sol.	A.CPU	#Sol.	A.CPU
Instances	55		627		397	
P_NI	37	172.76	254	152.04	233	131.32
P_NI+BCE	<b>41</b>	237.58	241	105.02	234	105.65
P_NI+BCE+RS	35	177.37	240	120.70	247	75.42
P_NI+RS	37	246.68	265	154.84	254	75.00
P	37	236.26	237	132.83	249	31.31
P+BCE	38	180.22	227	70.08	248	27.60
P+BCE+RS	37	209.48	221	85.36	259	32.19
P+RS	34	151.77	238	146.93	273	40.08
L	36	101.92	270	75.45	274	30.64
L+BCE	37	67.88	271	95.89	276	25.49
L+BCE+RS	38	96.02	271	73.90	275	15.90
L+RS	38	161.71	<b>276</b>	91.15	<b>289</b>	27.85
WMSU1	39	165.64	241	149.01	232	165.35

# To sum up ...

## Direct preprocessing

- ▶ plain-MUS: any clause elimination and BCP.
- ▶ group-MUS: only MUS-preserving (sp. monotone) clause elimination.
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## Experimental evaluation

Good results on on industrially-relevant group-MUS computation problems

Improvements on weighted partial MaxSAT problems

# Current and Future Work

- ▶ Analyze additional preprocessing techniques.
- ▶ Relax the conditions for labelled subsumption elimination.
- ▶ Proper implementation and additional MaxSAT algorithms.

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Thank you for your attention !

1. A. Belov, M. Jarvisalo and J. Marques-Silva. Formula Preprocessing in MUS Extraction. In Proc. TACAS 2013, 2013
2. A. Belov, A. Morgado, J. Marques-Silva. SAT-based Preprocessing for MaxSAT. In Proc. LPAR'19 (to appear), 2013.
3. A. Belov and J. Marques-Silva. Generalizing Redundancy in Propositional Logic: Foundations and Hitting Sets Duality. arXiv, 2012.