

# MUSes and MCSes

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# SAT

- Boolean satisfiability in Conjunctive Normal Form (CNF)
  - True/False variables  $x, y, z, \dots$
  - Literals  $x, \bar{x}, \dots$
  - Clauses: disjunctions (sets) of literals  $x \vee \bar{y} \vee z$
  - Formulas: conjunctions (sets) of clauses

# Minimal Correction Sets

- $F \setminus C$  is satisfiable, no larger subset of  $F$  is

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$$\begin{array}{ccc} (x_1) & (x_2) & (x_3) \\ (\bar{x}_1 \vee \bar{x}_2) & (\bar{x}_1 \vee \bar{x}_3) & (\bar{x}_2 \vee \bar{x}_3) \end{array}$$

# Minimal Unsatisfiable Sets

- $U \subseteq F$  is unsatisfiable, no smaller subset of  $F$  is

*Also called minimal cores*

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# MUS applications

- **Analysis of over-constrained systems** [Junker, 2004]
- **Diagnosis** [Han and Lee, 1999]
- **Type error Debugging** [Stuckey et al, 2003]
- **Function Decomposition** [Belov et al, 2012]
- **Approximate model counting** [Ivrii et al, 2016]
- **MaxSAT** [Morgado et al, 2013; Davies and Bacchus, 2011]

# MUS applications – over-constrained systems

- Help user find the bug in their model



# MUS applications – over-constrained systems

- Help user find the bug in their model
- Also for indirectly generated models
  - Compiler-generated model for type-checking

## MUS applications – Approximate model counting

- Add a system of large XOR constraints and do complete model counting
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- But SAT solvers generally poor at handling large XOR constraints
- Find “input” variables, restrict XORs to these
- MUSes of

$$F(\mathbf{X}) \wedge F(\mathbf{Y}) \wedge \bigwedge (x_i = y_i) \wedge \bigvee (x_i \neq y_i)$$

# MCS applications

- **MaxSAT** [Mencía et al, 2015; Bjorner and Narodytska, 2015]
- **Finding MUSes!** [Bacchus and K, 2015]

## Hitting set duality

$$\begin{array}{ccc} (x_1) & (x_2) & (x_3) \\ (\bar{x}_1 \vee \bar{x}_2) & (\bar{x}_1 \vee \bar{x}_3) & (\bar{x}_2 \vee \bar{x}_3) \end{array}$$

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# Hitting set duality

- Every (minimal) CS is a hitting set of all (minimal) USEs
- Every (minimal) US is a hitting set of all (minimal) CSEs

[Reiter, 1987]

# Interesting Queries

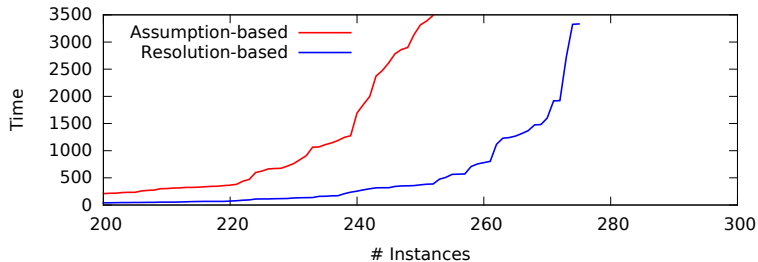
- Discover a single MCS/MUS
  - $D^P$ –complete [Liberatore, 2005]
- Enumerate all MCSes/MUSes
  - Hypergraph transversal
- Find minimum MCS/MUS
  - MUS:  $\Sigma_2^P$ –complete [Gupta, 2006]
  - MCS:  $D^P$ –complete (MaxSAT)



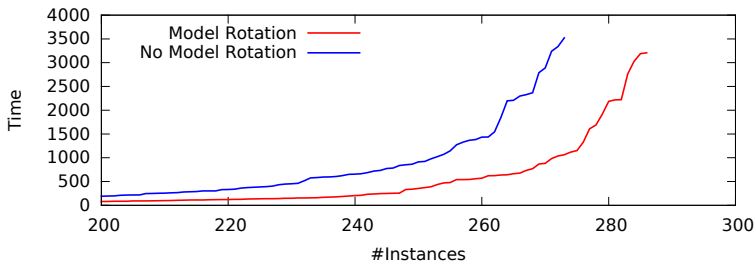
## Discovering MUSes

- Algorithm
- Incremental solver technology
- Avoiding calling the SAT solver
  - Trimming [Belov et al, 2014]
  - Model Rotation [Belov and Marques-Silva, 2011]
  - Clause set refinement [Nadel et al, 2014]

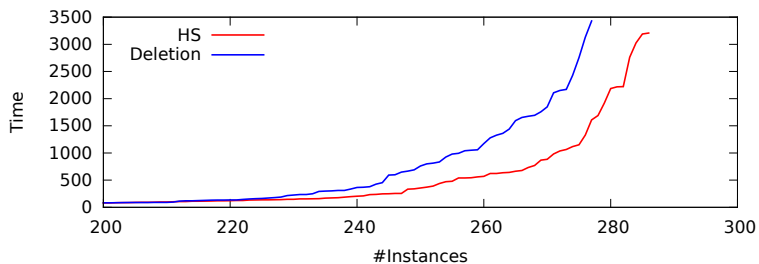
# Impact of incremental solver technology



# Impact of fewer SAT calls



# Impact of algorithm



# Algorithms

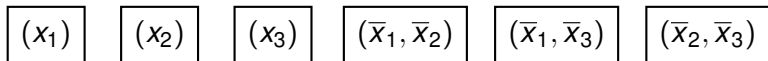
- **Deletion-based** [Belov and Marques-Silva, 2012]
- **Addition-based** [Gregoire et al, 2008; van Maaren and Wieringa, 2008]
- **Dichotomic** [Junker, 2004]
- **Exploit duality** [Bacchus and K, 2015]

# Algorithms

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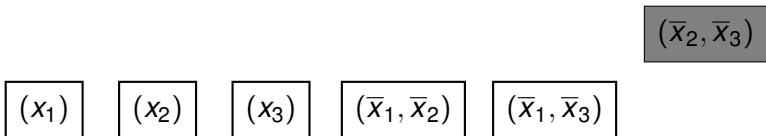
⇒ All applicable to any set minimization problem

## Deletion based

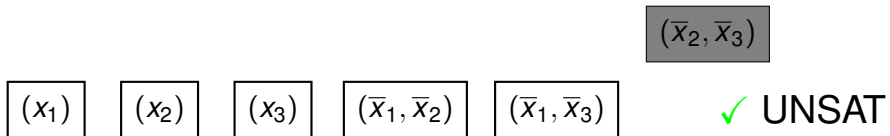




## Deletion based



## Deletion based



## Deletion based

$(\bar{x}_2, \bar{x}_3)$

$(x_1)$

$(x_2)$

$(x_3)$

$(\bar{x}_1, \bar{x}_2)$

$(\bar{x}_1, \bar{x}_3)$

## Deletion based

$(x_1)$     $(x_2)$     $(x_3)$     $(\bar{x}_1, \bar{x}_2)$

$(\bar{x}_1, \bar{x}_3)$

$(\bar{x}_2, \bar{x}_3)$

## Deletion based

$(x_1)$

$(x_2)$

$(x_3)$

$(\bar{x}_1, \bar{x}_2)$

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$(\bar{x}_2, \bar{x}_3)$

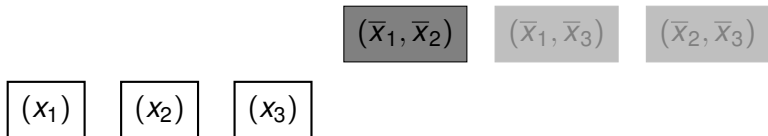
✓ UNSAT

## Deletion based

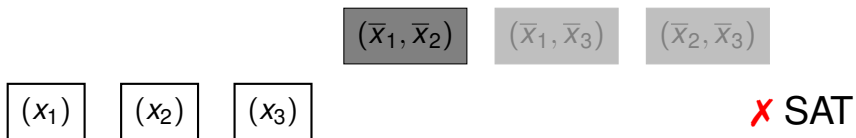
$(x_1)$     $(x_2)$     $(x_3)$     $(\bar{x}_1, \bar{x}_2)$

$(\bar{x}_1, \bar{x}_3)$     $(\bar{x}_2, \bar{x}_3)$

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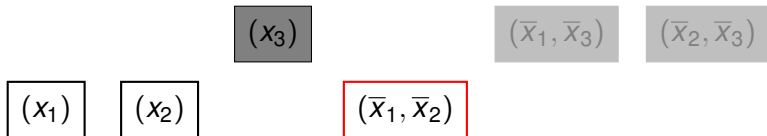


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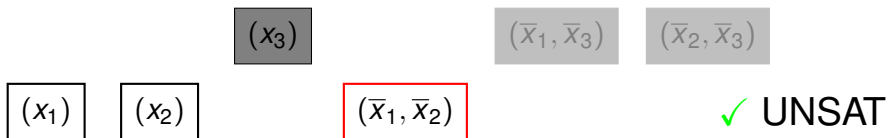
$(x_1)$     $(x_2)$     $(x_3)$     $(\bar{x}_1, \bar{x}_2)$

$(\bar{x}_1, \bar{x}_3)$     $(\bar{x}_2, \bar{x}_3)$

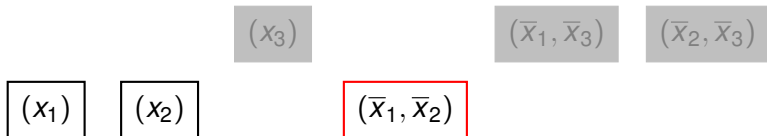
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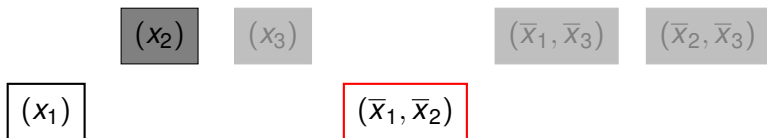
## Deletion based



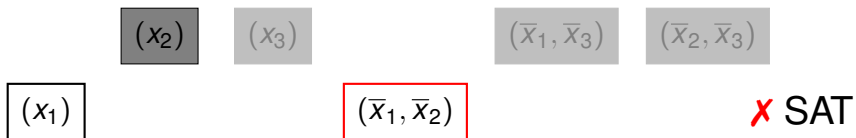
## Deletion based



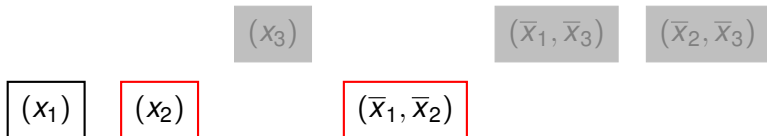
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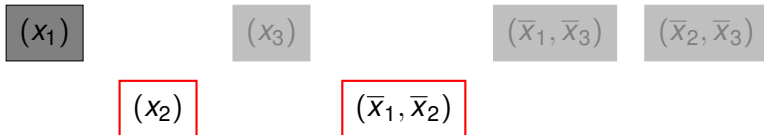
## Deletion based



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## Deletion based





## Deletion based

$(x_1)$

$(x_3)$

$(\bar{x}_1, \bar{x}_3)$

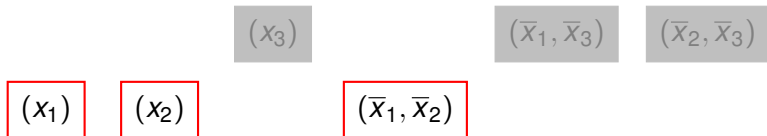
$(\bar{x}_2, \bar{x}_3)$

$(x_2)$

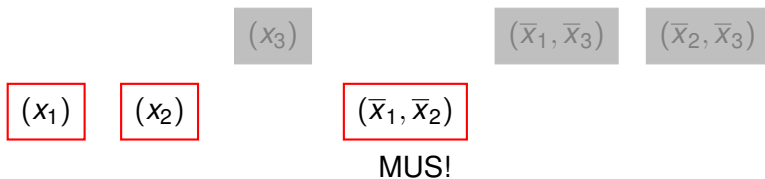
$(\bar{x}_1, \bar{x}_2)$

**X SAT**

## Deletion based



## Deletion based



## Exploiting Hitting Set duality

$(\bar{x}_1, \bar{x}_2)$

$(\bar{x}_1, \bar{x}_3)$

$(\bar{x}_2, \bar{x}_3)$

$(x_1)$

$(x_2)$

$(x_3)$

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$$(\bar{x}_2, \bar{x}_3)$$

$$(x_1)$$

$$(x_2)$$

$$(x_3)$$

**X** No MCS

## Exploiting Hitting Set duality

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$(\bar{x}_2, \bar{x}_3)$

$(x_1)$

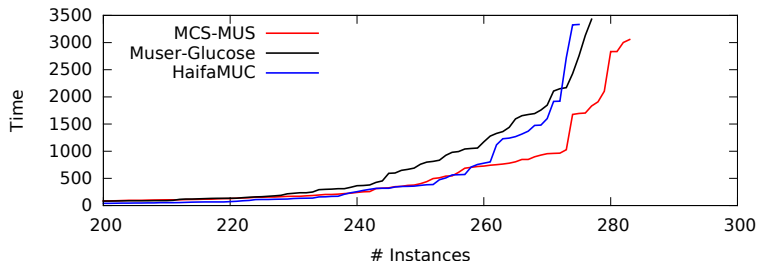
$(x_2)$

$(x_3)$

MUS

Is it faster to

- do  $m$  SAT checks?  
– or –
- discover  $k$  MCSes?



# The way forward

- Faster SAT solving 😊
  - Better use of existing SAT solvers
  - Extract more information from the solver
- Improved MCS extraction
- A new algorithm?

# More problems

- **Enumeration** [Liffiton and Sakallah, 2008; Liffiton et al, 2015; Bacchus and K, 2016]
- **Minimization** [Ignatiev et al, 2013; Ignatiev et al, 2015; Liffiton et al, 2009; Marques-Silva and Previti, 2014]
- **MCS extraction and enumeration** [O'Callaghan et al, 2005; Bailey and Stuckey, 2005; Felfernig et al, 2014; Nöhrer et al, 2012; Marques-Silva et al, 2013; Bacchus et al, 2014; Bacchus and K, 2015; Mencia et al, 2015]



Q?