

# Answer Set Solving in Practice

Torsten Schaub  
University of Potsdam  
torsten@cs.uni-potsdam.de



Potassco Slide Packages are licensed under a Creative Commons Attribution 3.0 Unported License.

# Conflict-driven ASP Solving: Overview

- 1 Motivation
- 2 Boolean constraints
- 3 Nogoods from logic programs
- 4 Conflict-driven nogood learning

# Outline

- 1 Motivation
- 2 Boolean constraints
- 3 Nogoods from logic programs
- 4 Conflict-driven nogood learning

# Motivation

- Goal Approach to computing stable models of logic programs, based on concepts from
  - Constraint Processing (CP) and
  - Satisfiability Testing (SAT)
- Idea View inferences in ASP as unit propagation on nogoods
- Benefits
  - A uniform constraint-based framework for different kinds of inferences in ASP
  - Advanced techniques from the areas of CP and SAT
  - Highly competitive implementation

# Outline

- 1 Motivation
- 2 Boolean constraints
- 3 Nogoods from logic programs
- 4 Conflict-driven nogood learning

# Assignments

- An assignment  $A$  over  $dom(A) = atom(P) \cup body(P)$  is a sequence

$$(\sigma_1, \dots, \sigma_n)$$

of signed literals  $\sigma_i$  of form  $\mathbf{T}v$  or  $\mathbf{F}v$  for  $v \in dom(A)$  and  $1 \leq i \leq n$

- $\mathbf{T}v$  expresses that  $v$  is *true* and  $\mathbf{F}v$  that it is *false*
- The complement,  $\bar{\sigma}$ , of a literal  $\sigma$  is defined as  $\overline{\mathbf{T}v} = \mathbf{F}v$  and  $\overline{\mathbf{F}v} = \mathbf{T}v$
- $A \circ \sigma$  stands for the result of appending  $\sigma$  to  $A$
- Given  $A = (\sigma_1, \dots, \sigma_{k-1}, \sigma_k, \dots, \sigma_n)$ , we let  $A[\sigma_k] = (\sigma_1, \dots, \sigma_{k-1})$
- We sometimes identify an assignment with the set of its literals
- Given this, we access *true* and *false* propositions in  $A$  via

$$A^T = \{v \in dom(A) \mid \mathbf{T}v \in A\} \text{ and } A^F = \{v \in dom(A) \mid \mathbf{F}v \in A\}$$

# Assignments

- An assignment  $A$  over  $dom(A) = atom(P) \cup body(P)$  is a sequence

$$(\sigma_1, \dots, \sigma_n)$$

of signed literals  $\sigma_i$  of form  $\mathbf{T}v$  or  $\mathbf{F}v$  for  $v \in dom(A)$  and  $1 \leq i \leq n$

- $\mathbf{T}v$  expresses that  $v$  is *true* and  $\mathbf{F}v$  that it is *false*
- The complement,  $\bar{\sigma}$ , of a literal  $\sigma$  is defined as  $\overline{\mathbf{T}v} = \mathbf{F}v$  and  $\overline{\mathbf{F}v} = \mathbf{T}v$
- $A \circ \sigma$  stands for the result of appending  $\sigma$  to  $A$
- Given  $A = (\sigma_1, \dots, \sigma_{k-1}, \sigma_k, \dots, \sigma_n)$ , we let  $A[\sigma_k] = (\sigma_1, \dots, \sigma_{k-1})$
- We sometimes identify an assignment with the set of its literals
- Given this, we access *true* and *false* propositions in  $A$  via

$$A^T = \{v \in dom(A) \mid \mathbf{T}v \in A\} \text{ and } A^F = \{v \in dom(A) \mid \mathbf{F}v \in A\}$$

# Assignments

- An assignment  $A$  over  $dom(A) = atom(P) \cup body(P)$  is a sequence

$$(\sigma_1, \dots, \sigma_n)$$

of signed literals  $\sigma_i$  of form  $Tv$  or  $Fv$  for  $v \in dom(A)$  and  $1 \leq i \leq n$

- $Tv$  expresses that  $v$  is *true* and  $Fv$  that it is *false*
- The complement,  $\bar{\sigma}$ , of a literal  $\sigma$  is defined as  $\overline{Tv} = Fv$  and  $\overline{Fv} = Tv$
- $A \circ \sigma$  stands for the result of appending  $\sigma$  to  $A$ 
  - Given  $A = (\sigma_1, \dots, \sigma_{k-1}, \sigma_k, \dots, \sigma_n)$ , we let  $A[\sigma_k] = (\sigma_1, \dots, \sigma_{k-1})$
  - We sometimes identify an assignment with the set of its literals
  - Given this, we access *true* and *false* propositions in  $A$  via

$$A^T = \{v \in dom(A) \mid Tv \in A\} \text{ and } A^F = \{v \in dom(A) \mid Fv \in A\}$$



# Assignments

- An assignment  $A$  over  $dom(A) = atom(P) \cup body(P)$  is a sequence

$$(\sigma_1, \dots, \sigma_n)$$

of signed literals  $\sigma_i$  of form  $Tv$  or  $Fv$  for  $v \in dom(A)$  and  $1 \leq i \leq n$

- $Tv$  expresses that  $v$  is *true* and  $Fv$  that it is *false*
- The complement,  $\bar{\sigma}$ , of a literal  $\sigma$  is defined as  $\overline{Tv} = Fv$  and  $\overline{Fv} = Tv$
- $A \circ \sigma$  stands for the result of appending  $\sigma$  to  $A$
- Given  $A = (\sigma_1, \dots, \sigma_{k-1}, \sigma_k, \dots, \sigma_n)$ , we let  $A[\sigma_k] = (\sigma_1, \dots, \sigma_{k-1})$
- We sometimes identify an assignment with the set of its literals
- Given this, we access *true* and *false* propositions in  $A$  via

$$A^T = \{v \in dom(A) \mid Tv \in A\} \text{ and } A^F = \{v \in dom(A) \mid Fv \in A\}$$

# Assignments

- An assignment  $A$  over  $dom(A) = atom(P) \cup body(P)$  is a sequence

$$(\sigma_1, \dots, \sigma_n)$$

of signed literals  $\sigma_i$  of form  $Tv$  or  $Fv$  for  $v \in dom(A)$  and  $1 \leq i \leq n$

- $Tv$  expresses that  $v$  is *true* and  $Fv$  that it is *false*
- The complement,  $\bar{\sigma}$ , of a literal  $\sigma$  is defined as  $\overline{Tv} = Fv$  and  $\overline{Fv} = Tv$
- $A \circ \sigma$  stands for the result of appending  $\sigma$  to  $A$
- Given  $A = (\sigma_1, \dots, \sigma_{k-1}, \sigma_k, \dots, \sigma_n)$ , we let  $A[\sigma_k] = (\sigma_1, \dots, \sigma_{k-1})$
- We sometimes identify an assignment with the set of its literals
- Given this, we access *true* and *false* propositions in  $A$  via

$$A^T = \{v \in dom(A) \mid Tv \in A\} \text{ and } A^F = \{v \in dom(A) \mid Fv \in A\}$$

# Assignments

- An assignment  $A$  over  $dom(A) = atom(P) \cup body(P)$  is a sequence

$$(\sigma_1, \dots, \sigma_n)$$

of signed literals  $\sigma_i$  of form  $\mathbf{T}v$  or  $\mathbf{F}v$  for  $v \in dom(A)$  and  $1 \leq i \leq n$

- $\mathbf{T}v$  expresses that  $v$  is *true* and  $\mathbf{F}v$  that it is *false*
- The complement,  $\bar{\sigma}$ , of a literal  $\sigma$  is defined as  $\overline{\mathbf{T}v} = \mathbf{F}v$  and  $\overline{\mathbf{F}v} = \mathbf{T}v$
- $A \circ \sigma$  stands for the result of appending  $\sigma$  to  $A$
- Given  $A = (\sigma_1, \dots, \sigma_{k-1}, \sigma_k, \dots, \sigma_n)$ , we let  $A[\sigma_k] = (\sigma_1, \dots, \sigma_{k-1})$
- We sometimes identify an assignment with the set of its literals
- Given this, we access *true* and *false* propositions in  $A$  via

$$A^T = \{v \in dom(A) \mid \mathbf{T}v \in A\} \text{ and } A^F = \{v \in dom(A) \mid \mathbf{F}v \in A\}$$

# Assignments

- An assignment  $A$  over  $dom(A) = atom(P) \cup body(P)$  is a sequence

$$(\sigma_1, \dots, \sigma_n)$$

of signed literals  $\sigma_i$  of form  $\mathbf{T}v$  or  $\mathbf{F}v$  for  $v \in dom(A)$  and  $1 \leq i \leq n$

- $\mathbf{T}v$  expresses that  $v$  is *true* and  $\mathbf{F}v$  that it is *false*
- The complement,  $\bar{\sigma}$ , of a literal  $\sigma$  is defined as  $\overline{\mathbf{T}v} = \mathbf{F}v$  and  $\overline{\mathbf{F}v} = \mathbf{T}v$
- $A \circ \sigma$  stands for the result of appending  $\sigma$  to  $A$
- Given  $A = (\sigma_1, \dots, \sigma_{k-1}, \sigma_k, \dots, \sigma_n)$ , we let  $A[\sigma_k] = (\sigma_1, \dots, \sigma_{k-1})$
- We sometimes identify an assignment with the set of its literals
- Given this, we access *true* and *false* propositions in  $A$  via

$$A^T = \{v \in dom(A) \mid \mathbf{T}v \in A\} \text{ and } A^F = \{v \in dom(A) \mid \mathbf{F}v \in A\}$$

## Nogoods, solutions, and unit propagation

- A **nogood** is a set  $\{\sigma_1, \dots, \sigma_n\}$  of signed literals, expressing a **constraint** violated by any assignment containing  $\sigma_1, \dots, \sigma_n$
- An assignment  $A$  such that  $A^T \cup A^F = \text{dom}(A)$  and  $A^T \cap A^F = \emptyset$  is a solution for a set  $\Delta$  of nogoods, if  $\delta \not\subseteq A$  for all  $\delta \in \Delta$
- For a nogood  $\delta$ , a literal  $\sigma \in \delta$ , and an assignment  $A$ , we say that  $\bar{\sigma}$  is unit-resulting for  $\delta$  wrt  $A$ , if
  - 1  $\delta \setminus A = \{\sigma\}$  and
  - 2  $\bar{\sigma} \notin A$
- For a set  $\Delta$  of nogoods and an assignment  $A$ , unit propagation is the iterated process of extending  $A$  with unit-resulting literals until no further literal is unit-resulting for any nogood in  $\Delta$

## Nogoods, solutions, and unit propagation

- A **nogood** is a set  $\{\sigma_1, \dots, \sigma_n\}$  of signed literals, expressing a **constraint** violated by any assignment containing  $\sigma_1, \dots, \sigma_n$
- An assignment  $A$  such that  $A^T \cup A^F = \text{dom}(A)$  and  $A^T \cap A^F = \emptyset$  is a **solution** for a set  $\Delta$  of nogoods, if  $\delta \not\subseteq A$  for all  $\delta \in \Delta$
- For a nogood  $\delta$ , a literal  $\sigma \in \delta$ , and an assignment  $A$ , we say that  $\bar{\sigma}$  is unit-resulting for  $\delta$  wrt  $A$ , if
  - 1  $\delta \setminus A = \{\sigma\}$  and
  - 2  $\bar{\sigma} \notin A$
- For a set  $\Delta$  of nogoods and an assignment  $A$ , unit propagation is the iterated process of extending  $A$  with unit-resulting literals until no further literal is unit-resulting for any nogood in  $\Delta$

## Nogoods, solutions, and unit propagation

- A **nogood** is a set  $\{\sigma_1, \dots, \sigma_n\}$  of signed literals, expressing a **constraint** violated by any assignment containing  $\sigma_1, \dots, \sigma_n$
- An assignment  $A$  such that  $A^T \cup A^F = \text{dom}(A)$  and  $A^T \cap A^F = \emptyset$  is a **solution** for a set  $\Delta$  of nogoods, if  $\delta \not\subseteq A$  for all  $\delta \in \Delta$
- For a nogood  $\delta$ , a literal  $\sigma \in \delta$ , and an assignment  $A$ , we say that  $\bar{\sigma}$  is **unit-resulting** for  $\delta$  wrt  $A$ , if
  - 1  $\delta \setminus A = \{\sigma\}$  and
  - 2  $\bar{\sigma} \notin A$
- For a set  $\Delta$  of nogoods and an assignment  $A$ , unit propagation is the iterated process of extending  $A$  with unit-resulting literals until no further literal is unit-resulting for any nogood in  $\Delta$

## Nogoods, solutions, and unit propagation

- A **nogood** is a set  $\{\sigma_1, \dots, \sigma_n\}$  of signed literals, expressing a **constraint** violated by any assignment containing  $\sigma_1, \dots, \sigma_n$
- An assignment  $A$  such that  $A^T \cup A^F = \text{dom}(A)$  and  $A^T \cap A^F = \emptyset$  is a **solution** for a set  $\Delta$  of nogoods, if  $\delta \not\subseteq A$  for all  $\delta \in \Delta$
- For a nogood  $\delta$ , a literal  $\sigma \in \delta$ , and an assignment  $A$ , we say that  $\bar{\sigma}$  is **unit-resulting** for  $\delta$  wrt  $A$ , if
  - 1  $\delta \setminus A = \{\sigma\}$  and
  - 2  $\bar{\sigma} \notin A$
- For a set  $\Delta$  of nogoods and an assignment  $A$ , **unit propagation** is the iterated process of extending  $A$  with unit-resulting literals until no further literal is unit-resulting for any nogood in  $\Delta$



# Outline

- 1 Motivation
- 2 Boolean constraints
- 3 Nogoods from logic programs
- 4 Conflict-driven nogood learning

# Outline

- 1 Motivation
- 2 Boolean constraints
- 3 Nogoods from logic programs
  - Nogoods from program completion
  - Nogoods from loop formulas
- 4 Conflict-driven nogood learning

# Nogoods from logic programs via program completion

The completion of a logic program  $P$  can be defined as follows:

$$\{v_B \leftrightarrow a_1 \wedge \cdots \wedge a_m \wedge \neg a_{m+1} \wedge \cdots \wedge \neg a_n \mid$$

$$B \in \text{body}(P) \text{ and } B = \{a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n\}\}$$

$$\cup \{a \leftrightarrow v_{B_1} \vee \cdots \vee v_{B_k} \mid$$

$$a \in \text{atom}(P) \text{ and } \text{body}_P(a) = \{B_1, \dots, B_k\}\},$$

where  $\text{body}_P(a) = \{\text{body}(r) \mid r \in P \text{ and } \text{head}(r) = a\}$

# Nogoods from logic programs via program completion

- The (body-oriented) equivalence

$$v_B \leftrightarrow a_1 \wedge \cdots \wedge a_m \wedge \neg a_{m+1} \wedge \cdots \wedge \neg a_n$$

can be decomposed into two implications:

# Nogoods from logic programs via program completion

- The (body-oriented) equivalence

$$v_B \leftrightarrow a_1 \wedge \cdots \wedge a_m \wedge \neg a_{m+1} \wedge \cdots \wedge \neg a_n$$

can be decomposed into two implications:

- 1  $v_B \rightarrow a_1 \wedge \cdots \wedge a_m \wedge \neg a_{m+1} \wedge \cdots \wedge \neg a_n$

is equivalent to the conjunction of

$$\neg v_B \vee a_1, \dots, \neg v_B \vee a_m, \neg v_B \vee \neg a_{m+1}, \dots, \neg v_B \vee \neg a_n$$

and induces the set of nogoods

$$\Delta(B) = \{ \{ \mathbf{TB}, \mathbf{Fa}_1 \}, \dots, \{ \mathbf{TB}, \mathbf{Fa}_m \}, \{ \mathbf{TB}, \mathbf{Ta}_{m+1} \}, \dots, \{ \mathbf{TB}, \mathbf{Ta}_n \} \}$$

# Nogoods from logic programs via program completion

## ■ The (body-oriented) equivalence

$$v_B \leftrightarrow a_1 \wedge \cdots \wedge a_m \wedge \neg a_{m+1} \wedge \cdots \wedge \neg a_n$$

can be decomposed into two implications:

$$\boxed{2} \quad a_1 \wedge \cdots \wedge a_m \wedge \neg a_{m+1} \wedge \cdots \wedge \neg a_n \rightarrow v_B$$

gives rise to the nogood

$$\delta(B) = \{\mathbf{FB}, \mathbf{Ta}_1, \dots, \mathbf{Ta}_m, \mathbf{Fa}_{m+1}, \dots, \mathbf{Fa}_n\}$$

# Nogoods from logic programs via program completion

- Analogously, the (atom-oriented) equivalence

$$a \leftrightarrow v_{B_1} \vee \cdots \vee v_{B_k}$$

yields the nogoods

**1**  $\Delta(a) = \{ \{ \mathbf{F}a, \mathbf{T}B_1 \}, \dots, \{ \mathbf{F}a, \mathbf{T}B_k \} \}$  and

**2**  $\delta(a) = \{ \mathbf{T}a, \mathbf{F}B_1, \dots, \mathbf{F}B_k \}$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{\mathbf{T}a, \mathbf{F}B_1, \dots, \mathbf{F}B_k\} \quad \text{and} \quad \{\{\mathbf{F}a, \mathbf{T}B_1\}, \dots, \{\mathbf{F}a, \mathbf{T}B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x \leftarrow y$	$\{\mathbf{T}x, \mathbf{F}\{y\}, \mathbf{F}\{\sim z\}\}$
$x \leftarrow \sim z$	$\{\{\mathbf{F}x, \mathbf{T}\{y\}\}, \{\mathbf{F}x, \mathbf{T}\{\sim z\}\}\}$

For nogood  $\{\mathbf{T}x, \mathbf{F}\{y\}, \mathbf{F}\{\sim z\}\}$ , the signed literal

$\mathbf{F}x$  is unit-resulting wrt assignment  $(\mathbf{F}\{y\}, \mathbf{F}\{\sim z\})$  and

$\mathbf{T}\{\sim z\}$  is unit-resulting wrt assignment  $(\mathbf{T}x, \mathbf{F}\{y\})$



# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{\mathbf{T}a, \mathbf{F}B_1, \dots, \mathbf{F}B_k\} \quad \text{and} \quad \{\{\mathbf{F}a, \mathbf{T}B_1\}, \dots, \{\mathbf{F}a, \mathbf{T}B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{\mathbf{T}x, \mathbf{F}\{y\}, \mathbf{F}\{\sim z\}\}$$

$$\{\{\mathbf{F}x, \mathbf{T}\{y\}\}, \{\mathbf{F}x, \mathbf{T}\{\sim z\}\}\}$$

For nogood  $\{\mathbf{T}x, \mathbf{F}\{y\}, \mathbf{F}\{\sim z\}\}$ , the signed literal

$\mathbf{F}x$  is unit-resulting wrt assignment  $(\mathbf{F}\{y\}, \mathbf{F}\{\sim z\})$  and

$\mathbf{T}\{\sim z\}$  is unit-resulting wrt assignment  $(\mathbf{T}x, \mathbf{F}\{y\})$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{\mathbf{T}a, \mathbf{F}B_1, \dots, \mathbf{F}B_k\} \quad \text{and} \quad \{\{\mathbf{F}a, \mathbf{T}B_1\}, \dots, \{\mathbf{F}a, \mathbf{T}B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x \leftarrow y$	$\{\mathbf{T}x, \mathbf{F}\{y\}, \mathbf{F}\{\sim z\}\}$
$x \leftarrow \sim z$	$\{\{\mathbf{F}x, \mathbf{T}\{y\}\}, \{\mathbf{F}x, \mathbf{T}\{\sim z\}\}\}$

For nogood  $\{\mathbf{T}x, \mathbf{F}\{y\}, \mathbf{F}\{\sim z\}\}$ , the signed literal

- $\mathbf{F}x$  is unit-resulting wrt assignment  $(\mathbf{F}\{y\}, \mathbf{F}\{\sim z\})$  and
- $\mathbf{T}\{\sim z\}$  is unit-resulting wrt assignment  $(\mathbf{T}x, \mathbf{F}\{y\})$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{T a, F B_1, \dots, F B_k\} \quad \text{and} \quad \{\{F a, T B_1\}, \dots, \{F a, T B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{y, \sim z\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{T x, F\{y\}, F\{\sim z\}\}$$

$$\{\{F x, T\{y\}\}, \{F x, T\{\sim z\}\}\}$$

For nogood  $\{T x, F\{y\}, F\{\sim z\}\}$ , the signed literal

- $F x$  is unit-resulting wrt assignment  $(F\{y\}, F\{\sim z\})$  and
- $T\{\sim z\}$  is unit-resulting wrt assignment  $(T x, F\{y\})$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{T a, F B_1, \dots, F B_k\} \quad \text{and} \quad \{\{F a, T B_1\}, \dots, \{F a, T B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{T x, F\{y\}, F\{\sim z\}\}$$

$$\{\{F x, T\{y\}\}, \{F x, T\{\sim z\}\}\}$$

For nogood  $\{T x, F\{y\}, F\{\sim z\}\}$ , the signed literal

- $F x$  is unit-resulting wrt assignment  $(F\{y\}, F\{\sim z\})$  and
- $T\{\sim z\}$  is unit-resulting wrt assignment  $(T x, F\{y\})$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{T a, F B_1, \dots, F B_k\} \quad \text{and} \quad \{\{F a, T B_1\}, \dots, \{F a, T B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{T x, F\{y\}, F\{\sim z\}\}$$

$$\{\{F x, T\{y\}\}, \{F x, T\{\sim z\}\}\}$$

For nogood  $\{T x, F\{y\}, F\{\sim z\}\}$ , the signed literal

- $F x$  is unit-resulting wrt assignment  $(F\{y\}, F\{\sim z\})$  and
- $T\{\sim z\}$  is unit-resulting wrt assignment  $(T x, F\{y\})$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{T a, F B_1, \dots, F B_k\} \quad \text{and} \quad \{\{F a, T B_1\}, \dots, \{F a, T B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{T x, F\{y\}, F\{\sim z\}\}$$

$$\{\{F x, T\{y\}\}, \{F x, T\{\sim z\}\}\}$$

For nogood  $\{T x, F\{y\}, F\{\sim z\}\}$ , the signed literal

- $F x$  is unit-resulting wrt assignment  $(F\{y\}, F\{\sim z\})$  and
- $T\{\sim z\}$  is unit-resulting wrt assignment  $(T x, F\{y\})$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{T a, F B_1, \dots, F B_k\} \quad \text{and} \quad \{\{F a, T B_1\}, \dots, \{F a, T B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{T x, F\{y\}, F\{\sim z\}\}$$

$$\{\{F x, T\{y\}\}, \{F x, T\{\sim z\}\}\}$$

For nogood  $\{T x, F\{y\}, F\{\sim z\}\}$ , the signed literal

- $F x$  is unit-resulting wrt assignment  $(F\{y\}, F\{\sim z\})$  and
- $T\{\sim z\}$  is unit-resulting wrt assignment  $(T x, F\{y\})$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{T a, F B_1, \dots, F B_k\} \quad \text{and} \quad \{\{F a, T B_1\}, \dots, \{F a, T B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{T x, F\{y\}, F\{\sim z\}\}$$

$$\{\{F x, T\{y\}\}, \{F x, T\{\sim z\}\}\}$$

For nogood  $\{T x, F\{y\}, F\{\sim z\}\}$ , the signed literal

- $F x$  is unit-resulting wrt assignment  $(F\{y\}, F\{\sim z\})$  and
- $T\{\sim z\}$  is unit-resulting wrt assignment  $(T x, F\{y\})$



# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{T a, F B_1, \dots, F B_k\} \quad \text{and} \quad \{\{F a, T B_1\}, \dots, \{F a, T B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{T x, F\{y\}, F\{\sim z\}\}$$

$$\{\{F x, T\{y\}\}, \{F x, T\{\sim z\}\}\}$$

For nogood  $\{T x, F\{y\}, F\{\sim z\}\}$ , the signed literal

- $F x$  is unit-resulting wrt assignment  $(F\{y\}, F\{\sim z\})$  and
- $T\{\sim z\}$  is unit-resulting wrt assignment  $(T x, F\{y\})$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{T a, F B_1, \dots, F B_k\} \quad \text{and} \quad \{\{F a, T B_1\}, \dots, \{F a, T B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{T x, F\{y\}, F\{\sim z\}\}$$

$$\{\{F x, T\{y\}\}, \{F x, T\{\sim z\}\}\}$$

For nogood  $\{T x, F\{y\}, F\{\sim z\}\}$ , the signed literal

- $F x$  is unit-resulting wrt assignment  $(F\{y\}, F\{\sim z\})$  and
- $T\{\sim z\}$  is unit-resulting wrt assignment  $(T x, F\{y\})$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{T a, F B_1, \dots, F B_k\} \quad \text{and} \quad \{\{F a, T B_1\}, \dots, \{F a, T B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{T x, F\{y\}, F\{\sim z\}\}$$

$$\{\{F x, T\{y\}\}, \{F x, T\{\sim z\}\}\}$$

For nogood  $\{T x, F\{y\}, F\{\sim z\}\}$ , the signed literal

- $F x$  is unit-resulting wrt assignment  $(F\{y\}, F\{\sim z\})$  and
- $T\{\sim z\}$  is unit-resulting wrt assignment  $(T x, F\{y\})$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{T a, F B_1, \dots, F B_k\} \quad \text{and} \quad \{\{F a, T B_1\}, \dots, \{F a, T B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{\{y\}, \{\sim z\}\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{T x, F\{y\}, F\{\sim z\}\}$$

$$\{\{F x, T\{y\}\}, \{F x, T\{\sim z\}\}\}$$

For nogood  $\{T x, F\{y\}, F\{\sim z\}\}$ , the signed literal

- $F x$  is unit-resulting wrt assignment  $(F\{y\}, F\{\sim z\})$  and
- $T\{\sim z\}$  is unit-resulting wrt assignment  $(T x, F\{y\})$

# Nogoods from logic programs

## atom-oriented nogoods

- For an atom  $a$  where  $body_P(a) = \{B_1, \dots, B_k\}$ , we get

$$\{T a, F B_1, \dots, F B_k\} \quad \text{and} \quad \{\{F a, T B_1\}, \dots, \{F a, T B_k\}\}$$

- Example Given Atom  $x$  with  $body(x) = \{y, \sim z\}$ , we obtain

$x$	$\leftarrow$	$y$
$x$	$\leftarrow$	$\sim z$

$$\{T x, F\{y\}, F\{\sim z\}\}$$

$$\{\{F x, T\{y\}\}, \{F x, T\{\sim z\}\}\}$$

For nogood  $\{T x, F\{y\}, F\{\sim z\}\}$ , the signed literal

- $F x$  is unit-resulting wrt assignment  $(F\{y\}, F\{\sim z\})$  and
- $T\{\sim z\}$  is unit-resulting wrt assignment  $(T x, F\{y\})$

# Nogoods from logic programs

## body-oriented nogoods

- For a body  $B = \{a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n\}$ , we get

$$\{FB, Ta_1, \dots, Ta_m, Fa_{m+1}, \dots, Fa_n\}$$

$$\{\{TB, Fa_1\}, \dots, \{TB, Fa_m\}, \{TB, Ta_{m+1}\}, \dots, \{TB, Ta_n\}\}$$

- Example Given Body  $\{x, \sim y\}$ , we obtain

$\dots \leftarrow x, \sim y$ $\vdots$ $\dots \leftarrow x, \sim y$	$\{F\{x, \sim y\}, Tx, Fy\}$ $\{\{T\{x, \sim y\}, Fx\}, \{T\{x, \sim y\}, Ty\}\}$
--	---

For nogood  $\delta(\{x, \sim y\}) = \{F\{x, \sim y\}, Tx, Fy\}$ , the signed literal

- $T\{x, \sim y\}$  is unit-resulting wrt assignment  $(Tx, Fy)$  and
- $Ty$  is unit-resulting wrt assignment  $(F\{x, \sim y\}, Tx)$

# Nogoods from logic programs

## body-oriented nogoods

- For a body  $B = \{a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n\}$ , we get

$$\{FB, Ta_1, \dots, Ta_m, Fa_{m+1}, \dots, Fa_n\}$$

$$\{\{TB, Fa_1\}, \dots, \{TB, Fa_m\}, \{TB, Ta_{m+1}\}, \dots, \{TB, Ta_n\}\}$$

- Example Given Body  $\{x, \sim y\}$ , we obtain

$$\boxed{\begin{array}{l} \dots \leftarrow x, \sim y \\ \vdots \\ \dots \leftarrow x, \sim y \end{array}}$$

$$\{F\{x, \sim y\}, Tx, Fy\}$$

$$\{\{T\{x, \sim y\}, Fx\}, \{T\{x, \sim y\}, Ty\}\}$$

For nogood  $\delta(\{x, \sim y\}) = \{F\{x, \sim y\}, Tx, Fy\}$ , the signed literal

- $T\{x, \sim y\}$  is unit-resulting wrt assignment  $(Tx, Fy)$  and
- $Ty$  is unit-resulting wrt assignment  $(F\{x, \sim y\}, Tx)$

# Nogoods from logic programs

## body-oriented nogoods

- For a body  $B = \{a_1, \dots, a_m, \sim a_{m+1}, \dots, \sim a_n\}$ , we get

$$\{FB, Ta_1, \dots, Ta_m, Fa_{m+1}, \dots, Fa_n\}$$

$$\{\{TB, Fa_1\}, \dots, \{TB, Fa_m\}, \{TB, Ta_{m+1}\}, \dots, \{TB, Ta_n\}\}$$

- Example Given Body  $\{x, \sim y\}$ , we obtain

$$\boxed{\begin{array}{l} \dots \leftarrow x, \sim y \\ \vdots \\ \dots \leftarrow x, \sim y \end{array}}$$

$$\{F\{x, \sim y\}, Tx, Fy\}$$

$$\{\{T\{x, \sim y\}, Fx\}, \{T\{x, \sim y\}, Ty\}\}$$

For nogood  $\delta(\{x, \sim y\}) = \{F\{x, \sim y\}, Tx, Fy\}$ , the signed literal

- $T\{x, \sim y\}$  is unit-resulting wrt assignment  $(Tx, Fy)$  and
- $Ty$  is unit-resulting wrt assignment  $(F\{x, \sim y\}, Tx)$



# Characterization of stable models

for tight logic programs

Let  $P$  be a logic program and

$$\begin{aligned} \Delta_P = & \quad \{\delta(a) \mid a \in \mathit{atom}(P)\} \cup \{\delta \in \Delta(a) \mid a \in \mathit{atom}(P)\} \\ & \cup \{\delta(B) \mid B \in \mathit{body}(P)\} \cup \{\delta \in \Delta(B) \mid B \in \mathit{body}(P)\} \end{aligned}$$

Theorem

Let  $P$  be a tight logic program. Then,

$X \subseteq \mathit{atom}(P)$  is a stable model of  $P$  iff

$X = A^T \cap \mathit{atom}(P)$  for a (unique) solution  $A$  for  $\Delta_P$

# Characterization of stable models

for tight logic programs

Let  $P$  be a logic program and

$$\begin{aligned} \Delta_P = & \quad \{\delta(a) \mid a \in \text{atom}(P)\} \cup \{\delta \in \Delta(a) \mid a \in \text{atom}(P)\} \\ & \cup \{\delta(B) \mid B \in \text{body}(P)\} \cup \{\delta \in \Delta(B) \mid B \in \text{body}(P)\} \end{aligned}$$

## Theorem

Let  $P$  be a *tight* logic program. Then,

$X \subseteq \text{atom}(P)$  is a stable model of  $P$  iff

$X = A^T \cap \text{atom}(P)$  for a (unique) solution  $A$  for  $\Delta_P$

# Characterization of stable models

for **tight** logic programs, ie. **free of positive recursion**

Let  $P$  be a logic program and

$$\begin{aligned} \Delta_P = & \quad \{\delta(a) \mid a \in \text{atom}(P)\} \cup \{\delta \in \Delta(a) \mid a \in \text{atom}(P)\} \\ & \cup \quad \{\delta(B) \mid B \in \text{body}(P)\} \cup \{\delta \in \Delta(B) \mid B \in \text{body}(P)\} \end{aligned}$$

## Theorem

Let  $P$  be a **tight** logic program. Then,

$X \subseteq \text{atom}(P)$  is a stable model of  $P$  iff

$X = A^T \cap \text{atom}(P)$  for a (unique) solution  $A$  for  $\Delta_P$

# Outline

- 1 Motivation
- 2 Boolean constraints
- 3 Nogoods from logic programs
  - Nogoods from program completion
  - Nogoods from loop formulas
- 4 Conflict-driven nogood learning

# Nogoods from logic programs

## via loop formulas

Let  $P$  be a normal logic program and recall that:

- For  $L \subseteq \text{atom}(P)$ , the external supports of  $L$  for  $P$  are

$$ES_P(L) = \{r \in P \mid \text{head}(r) \in L \text{ and } \text{body}(r)^+ \cap L = \emptyset\}$$

- The (disjunctive) loop formula of  $L$  for  $P$  is

$$\begin{aligned} LF_P(L) &= \left( \bigvee_{A \in L} A \right) \rightarrow \left( \bigvee_{r \in ES_P(L)} \text{body}(r) \right) \\ &\Leftrightarrow \left( \bigwedge_{r \in ES_P(L)} \neg \text{body}(r) \right) \rightarrow \left( \bigwedge_{A \in L} \neg A \right) \end{aligned}$$

- Note The loop formula of  $L$  enforces all atoms in  $L$  to be *false* whenever  $L$  is not externally supported

- The external bodies of  $L$  for  $P$  are

$$EB_P(L) = \{\text{body}(r) \mid r \in ES_P(L)\}$$

# Nogoods from logic programs

## via loop formulas

Let  $P$  be a normal logic program and recall that:

- For  $L \subseteq \text{atom}(P)$ , the external supports of  $L$  for  $P$  are

$$ES_P(L) = \{r \in P \mid \text{head}(r) \in L \text{ and } \text{body}(r)^+ \cap L = \emptyset\}$$

- The (disjunctive) loop formula of  $L$  for  $P$  is

$$\begin{aligned} LF_P(L) &= (\bigvee_{A \in L} A) \rightarrow (\bigvee_{r \in ES_P(L)} \text{body}(r)) \\ &\leftrightarrow (\bigwedge_{r \in ES_P(L)} \neg \text{body}(r)) \rightarrow (\bigwedge_{A \in L} \neg A) \end{aligned}$$

- Note The loop formula of  $L$  enforces all atoms in  $L$  to be *false* whenever  $L$  is not externally supported

- The external bodies of  $L$  for  $P$  are

$$EB_P(L) = \{\text{body}(r) \mid r \in ES_P(L)\}$$

# Nogoods from logic programs

## via loop formulas

Let  $P$  be a normal logic program and recall that:

- For  $L \subseteq \text{atom}(P)$ , the external supports of  $L$  for  $P$  are

$$ES_P(L) = \{r \in P \mid \text{head}(r) \in L \text{ and } \text{body}(r)^+ \cap L = \emptyset\}$$

- The (disjunctive) loop formula of  $L$  for  $P$  is

$$\begin{aligned} LF_P(L) &= (\bigvee_{A \in L} A) \rightarrow (\bigvee_{r \in ES_P(L)} \text{body}(r)) \\ &\Leftrightarrow (\bigwedge_{r \in ES_P(L)} \neg \text{body}(r)) \rightarrow (\bigwedge_{A \in L} \neg A) \end{aligned}$$

- Note The loop formula of  $L$  enforces all atoms in  $L$  to be *false* whenever  $L$  is not externally supported
- The external bodies of  $L$  for  $P$  are

$$EB_P(L) = \{\text{body}(r) \mid r \in ES_P(L)\}$$

# Nogoods from logic programs

## loop nogoods

- For a logic program  $P$  and some  $\emptyset \subset U \subseteq \text{atom}(P)$ , define the **loop nogood** of an atom  $a \in U$  as

$$\lambda(a, U) = \{ \mathbf{T}a, \mathbf{F}B_1, \dots, \mathbf{F}B_k \}$$

where  $EB_P(U) = \{B_1, \dots, B_k\}$

- We get the following set of loop nogoods for  $P$ :

$$\Lambda_P = \bigcup_{\emptyset \subset U \subseteq \text{atom}(P)} \{ \lambda(a, U) \mid a \in U \}$$

- The set  $\Lambda_P$  of loop nogoods denies cyclic support among *true* atoms



# Nogoods from logic programs

## loop nogoods

- For a logic program  $P$  and some  $\emptyset \subset U \subseteq \text{atom}(P)$ , define the **loop nogood** of an atom  $a \in U$  as

$$\lambda(a, U) = \{ \mathbf{T}a, \mathbf{F}B_1, \dots, \mathbf{F}B_k \}$$

where  $EB_P(U) = \{B_1, \dots, B_k\}$

- We get the following set of loop nogoods for  $P$ :

$$\Lambda_P = \bigcup_{\emptyset \subset U \subseteq \text{atom}(P)} \{ \lambda(a, U) \mid a \in U \}$$

- The set  $\Lambda_P$  of loop nogoods denies cyclic support among *true* atoms

# Nogoods from logic programs

## loop nogoods

- For a logic program  $P$  and some  $\emptyset \subset U \subseteq \text{atom}(P)$ , define the **loop nogood** of an atom  $a \in U$  as

$$\lambda(a, U) = \{ \mathbf{T}a, \mathbf{F}B_1, \dots, \mathbf{F}B_k \}$$

where  $EB_P(U) = \{B_1, \dots, B_k\}$

- We get the following set of loop nogoods for  $P$ :

$$\Lambda_P = \bigcup_{\emptyset \subset U \subseteq \text{atom}(P)} \{ \lambda(a, U) \mid a \in U \}$$

- The set  $\Lambda_P$  of loop nogoods denies cyclic support among *true* atoms

# Example

- Consider the program

$$\left\{ \begin{array}{ll} x \leftarrow \sim y & u \leftarrow x \\ y \leftarrow \sim x & u \leftarrow v \\ & v \leftarrow u, y \end{array} \right\}$$

- For  $u$  in the set  $\{u, v\}$ , we obtain the loop nogood:

$$\lambda(u, \{u, v\}) = \{\mathbf{T}u, \mathbf{F}\{x\}\}$$

Similarly for  $v$  in  $\{u, v\}$ , we get:

$$\lambda(v, \{u, v\}) = \{\mathbf{T}v, \mathbf{F}\{x\}\}$$

# Example

- Consider the program

$$\left\{ \begin{array}{ll} x \leftarrow \sim y & u \leftarrow x \\ y \leftarrow \sim x & u \leftarrow v \\ & v \leftarrow u, y \end{array} \right\}$$

- For  $u$  in the set  $\{u, v\}$ , we obtain the loop nogood:

$$\lambda(u, \{u, v\}) = \{\mathbf{T}u, \mathbf{F}\{x\}\}$$

Similarly for  $v$  in  $\{u, v\}$ , we get:

$$\lambda(v, \{u, v\}) = \{\mathbf{T}v, \mathbf{F}\{x\}\}$$

# Example

- Consider the program

$$\left\{ \begin{array}{ll} x \leftarrow \sim y & u \leftarrow x \\ y \leftarrow \sim x & u \leftarrow v \\ & v \leftarrow u, y \end{array} \right\}$$

- For  $u$  in the set  $\{u, v\}$ , we obtain the loop nogood:

$$\lambda(u, \{u, v\}) = \{\mathbf{T}u, \mathbf{F}\{x\}\}$$

Similarly for  $v$  in  $\{u, v\}$ , we get:

$$\lambda(v, \{u, v\}) = \{\mathbf{T}v, \mathbf{F}\{x\}\}$$

# Characterization of stable models

## Theorem

*Let  $P$  be a logic program. Then,*

*$X \subseteq \text{atom}(P)$  is a stable model of  $P$  iff*

*$X = A^T \cap \text{atom}(P)$  for a (unique) solution  $A$  for  $\Delta_P \cup \Lambda_P$*

## Some remarks

- Nogoods in  $\Lambda_P$  augment  $\Delta_P$  with conditions checking for unfounded sets, in particular, those being loops
- While  $|\Delta_P|$  is linear in the size of  $P$ ,  $\Lambda_P$  may contain exponentially many (non-redundant) loop nogoods

# Characterization of stable models

## Theorem

Let  $P$  be a logic program. Then,

$X \subseteq \text{atom}(P)$  is a stable model of  $P$  iff

$X = A^T \cap \text{atom}(P)$  for a (unique) solution  $A$  for  $\Delta_P \cup \Lambda_P$

## Some remarks

- Nogoods in  $\Lambda_P$  augment  $\Delta_P$  with conditions checking for **unfounded sets**, in particular, those being loops
- While  $|\Delta_P|$  is linear in the size of  $P$ ,  $\Lambda_P$  may contain **exponentially many** (non-redundant) loop nogoods

# Outline

- 1 Motivation
- 2 Boolean constraints
- 3 Nogoods from logic programs
- 4 Conflict-driven nogood learning



# Towards conflict-driven search

Boolean constraint solving algorithms pioneered for SAT led to:

- Traditional DPLL-style approach  
(DPLL stands for 'Davis-Putnam-Logemann-Loveland')
  - (Unit) propagation
  - (Chronological) backtracking
  - in ASP, eg *smodels*
- Modern CDCL-style approach  
(CDCL stands for 'Conflict-Driven Constraint Learning')
  - (Unit) propagation
  - Conflict analysis (via resolution)
  - Learning + Backjumping + Assertion
  - in ASP, eg *clasp*

## DPLL-style solving

**loop**

```

propagate                                // deterministically assign literals
if no conflict then
    if all variables assigned then return solution
    else decide                            // non-deterministically assign some literal
else
    if top-level conflict then return unsatisfiable
    else
        backtrack                          // unassign literals propagated after last decision
        flip                                // assign complement of last decision literal

```

## CDCL-style solving

**loop**

```

propagate                                // deterministically assign literals
if no conflict then
    if all variables assigned then return solution
    else decide                            // non-deterministically assign some literal
else
    if top-level conflict then return unsatisfiable
    else
        analyze                            // analyze conflict and add conflict constraint
        backjump                          // unassign literals until conflict constraint is unit

```

# Outline

- 1 Motivation
- 2 Boolean constraints
- 3 Nogoods from logic programs
- 4 Conflict-driven nogood learning
  - CDNL-ASP Algorithm
  - Nogood Propagation
  - Conflict Analysis

# Outline of CDNL-ASP algorithm

- Keep track of deterministic consequences by unit propagation on:
  - Program completion  $[\Delta_P]$
  - Loop nogoods, determined and recorded on demand  $[\Lambda_P]$
  - Dynamic nogoods, derived from conflicts and unfounded sets  $[\nabla]$
- When a nogood in  $\Delta_P \cup \nabla$  becomes violated:
  - Analyze the conflict by resolution  
(until reaching a Unique Implication Point, short: UIP)
  - Learn the derived conflict nogood  $\delta$
  - Backjump to the earliest (heuristic) choice such that the complement of the UIP is unit-resulting for  $\delta$
  - Assert the complement of the UIP and proceed  
(by unit propagation)
- Terminate when either:
  - Finding a stable model (a solution for  $\Delta_P \cup \Lambda_P$ )
  - Deriving a conflict independently of (heuristic) choices

# Outline of CDNL-ASP algorithm

- Keep track of deterministic consequences by unit propagation on:
  - Program completion  $[\Delta_P]$
  - Loop nogoods, determined and recorded on demand  $[\wedge_P]$
  - Dynamic nogoods, derived from conflicts and unfounded sets  $[\nabla]$
- When a nogood in  $\Delta_P \cup \nabla$  becomes **violated**:
  - **Analyze** the conflict by resolution  
(until reaching a Unique Implication Point, short: UIP)
  - **Learn** the derived conflict nogood  $\delta$
  - **Backjump** to the earliest (heuristic) choice such that the complement of the UIP is unit-resulting for  $\delta$
  - **Assert** the complement of the UIP and proceed  
(by unit propagation)
- Terminate when either:
  - Finding a stable model (a solution for  $\Delta_P \cup \wedge_P$ )
  - Deriving a conflict independently of (heuristic) choices

## Outline of CDNL-ASP algorithm

- Keep track of deterministic consequences by unit propagation on:
  - Program completion  $[\Delta_P]$
  - Loop nogoods, determined and recorded on demand  $[\Lambda_P]$
  - Dynamic nogoods, derived from conflicts and unfounded sets  $[\nabla]$
- When a nogood in  $\Delta_P \cup \nabla$  becomes violated:
  - Analyze the conflict by resolution  
(until reaching a Unique Implication Point, short: UIP)
  - Learn the derived conflict nogood  $\delta$
  - Backjump to the earliest (heuristic) choice such that the complement of the UIP is unit-resulting for  $\delta$
  - Assert the complement of the UIP and proceed  
(by unit propagation)
- Terminate when either:
  - Finding a stable model (a solution for  $\Delta_P \cup \Lambda_P$ )
  - Deriving a conflict independently of (heuristic) choices

## Algorithm 2: CDNL-ASP

```

Input      : A normal program  $P$ 
Output    : A stable model of  $P$  or “no stable model”

 $A := \emptyset$                                 // assignment over  $\text{atom}(P) \cup \text{body}(P)$ 
 $\nabla := \emptyset$                                // set of recorded nogoods
 $dl := 0$                                        // decision level

loop
   $(A, \nabla) := \text{NOGOODPROPAGATION}(P, \nabla, A)$ 
  if  $\varepsilon \subseteq A$  for some  $\varepsilon \in \Delta_P \cup \nabla$  then // conflict
    if  $\max(\{dlevel(\sigma) \mid \sigma \in \varepsilon\} \cup \{0\}) = 0$  then return no stable model
     $(\delta, dl) := \text{CONFLICTANALYSIS}(\varepsilon, P, \nabla, A)$ 
     $\nabla := \nabla \cup \{\delta\}$  // (temporarily) record conflict nogood
     $A := A \setminus \{\sigma \in A \mid dl < dlevel(\sigma)\}$  // backjumping
  else if  $A^T \cup A^F = \text{atom}(P) \cup \text{body}(P)$  then // stable model
    return  $A^T \cap \text{atom}(P)$ 
  else
     $\sigma_d := \text{SELECT}(P, \nabla, A)$  // decision
     $dl := dl + 1$ 
     $dlevel(\sigma_d) := dl$ 
     $A := A \circ \sigma_d$ 

```



## Observations

- Decision level  $dl$ , initially set to 0, is used to count the number of heuristically chosen literals in assignment  $A$
- For a heuristically chosen literal  $\sigma_d = \mathbf{T}a$  or  $\sigma_d = \mathbf{F}a$ , respectively, we require  $a \in (atom(P) \cup body(P)) \setminus (A^T \cup A^F)$
- For any literal  $\sigma \in A$ ,  $dl(\sigma)$  denotes the decision level of  $\sigma$ , viz. the value  $dl$  had when  $\sigma$  was assigned
- A conflict is detected from violation of a nogood  $\varepsilon \subseteq \Delta_P \cup \nabla$
- A conflict at decision level 0 (where  $A$  contains no heuristically chosen literals) indicates non-existence of stable models
- A nogood  $\delta$  derived by conflict analysis is asserting, that is, some literal is unit-resulting for  $\delta$  at a decision level  $k < dl$ 
  - After learning  $\delta$  and backjumping to decision level  $k$ , at least one literal is newly derivable by unit propagation
  - No explicit flipping of heuristically chosen literals !

## Observations

- Decision level  $dl$ , initially set to 0, is used to count the number of heuristically chosen literals in assignment  $A$
- For a heuristically chosen literal  $\sigma_d = \mathbf{T}a$  or  $\sigma_d = \mathbf{F}a$ , respectively, we require  $a \in (atom(P) \cup body(P)) \setminus (A^T \cup A^F)$
- For any literal  $\sigma \in A$ ,  $dl(\sigma)$  denotes the decision level of  $\sigma$ , viz. the value  $dl$  had when  $\sigma$  was assigned
- A conflict is detected from violation of a nogood  $\varepsilon \subseteq \Delta_P \cup \nabla$
- A conflict at decision level 0 (where  $A$  contains no heuristically chosen literals) indicates non-existence of stable models
- A nogood  $\delta$  derived by conflict analysis is **asserting**, that is, some literal is unit-resulting for  $\delta$  at a decision level  $k < dl$ 
  - After learning  $\delta$  and backjumping to decision level  $k$ , at least one literal is newly derivable by unit propagation
  - No explicit flipping of heuristically chosen literals !

## Observations

- Decision level  $dl$ , initially set to 0, is used to count the number of heuristically chosen literals in assignment  $A$
- For a heuristically chosen literal  $\sigma_d = \mathbf{T}a$  or  $\sigma_d = \mathbf{F}a$ , respectively, we require  $a \in (atom(P) \cup body(P)) \setminus (A^T \cup A^F)$
- For any literal  $\sigma \in A$ ,  $dl(\sigma)$  denotes the decision level of  $\sigma$ , viz. the value  $dl$  had when  $\sigma$  was assigned
- A conflict is detected from violation of a nogood  $\varepsilon \subseteq \Delta_P \cup \nabla$
- A conflict at decision level 0 (where  $A$  contains no heuristically chosen literals) indicates non-existence of stable models
- A nogood  $\delta$  derived by conflict analysis is **asserting**, that is, some literal is unit-resulting for  $\delta$  at a decision level  $k < dl$ 
  - After learning  $\delta$  and backjumping to decision level  $k$ , at least one literal is newly derivable by unit propagation
  - No explicit flipping of heuristically chosen literals !

## Example: CDNL-ASP

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $\vdots$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\vdots$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

## Example: CDNL-ASP

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $\vdots$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\vdots$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

# Example: CDNL-ASP

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $\vdots$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\vdots$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

## Example: CDNL-ASP

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $\vdots$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\vdots$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

## Example: CDNL-ASP

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $\vdots$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\vdots$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$



## Example: CDNL-ASP

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $\vdots$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\vdots$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

## Example: CDNL-ASP

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $\vdots$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\vdots$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

## Example: CDNL-ASP

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$	$Tx$ $\vdots$ $Tv$ $Fy$ $Fw$	$\{Tu, Fx\} \in \nabla$ $\vdots$ $\{Fv, T\{x\}\} \in \Delta(v)$ $\{Ty, F\{\sim x\}\} = \delta(y)$ $\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$

## Example: CDNL-ASP

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$	$Tx$ $\vdots$ $Tv$ $Fy$ $Fw$	$\{Tu, Fx\} \in \nabla$ $\vdots$ $\{Fv, T\{x\}\} \in \Delta(v)$ $\{Ty, F\{\sim x\}\} = \delta(y)$ $\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$

## Example: CDNL-ASP

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$	$Tx$ $\vdots$ $Tv$ $Fy$ $Fw$	$\{Tu, Fx\} \in \nabla$ $\vdots$ $\{Fv, T\{x\}\} \in \Delta(v)$ $\{Ty, F\{\sim x\}\} = \delta(y)$ $\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$

## Example: CDNL-ASP

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	<b><math>Tu</math></b>	$Tx$ $\vdots$ $Tv$ $Fy$ $Fw$	$\{Tu, Fx\} \in \nabla$ $\vdots$ $\{Fv, T\{x\}\} \in \Delta(v)$ $\{Ty, F\{\sim x\}\} = \delta(y)$ $\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$

# Outline

- 1 Motivation
- 2 Boolean constraints
- 3 Nogoods from logic programs
- 4 Conflict-driven nogood learning
  - CDNL-ASP Algorithm
  - Nogood Propagation
  - Conflict Analysis

# Outline of NogoodPropagation

- Derive deterministic consequences via:
  - Unit propagation on  $\Delta_P$  and  $\nabla$ ;
  - Unfounded sets  $U \subseteq atom(P)$
- Note that  $U$  is **unfounded** if  $EB_P(U) \subseteq A^F$ 
  - Note For any  $a \in U$ , we have  $(\lambda(a, U) \setminus \{Ta\}) \subseteq A$
- An “interesting” unfounded set  $U$  satisfies:

$$\emptyset \subset U \subseteq (atom(P) \setminus A^F)$$

- Wrt a fixpoint of unit propagation, such an unfounded set contains some loop of  $P$ 
  - Note Tight programs do not yield “interesting” unfounded sets !
- Given an unfounded set  $U$  and some  $a \in U$ , adding  $\lambda(a, U)$  to  $\nabla$  triggers a conflict or further derivations by unit propagation
  - Note Add loop nogoods atom by atom to eventually falsify all  $a \in U$



# Outline of NogoodPropagation

- Derive deterministic consequences via:
  - Unit propagation on  $\Delta_P$  and  $\nabla$ ;
  - Unfounded sets  $U \subseteq atom(P)$
- Note that  $U$  is **unfounded** if  $EB_P(U) \subseteq A^F$ 
  - Note For any  $a \in U$ , we have  $(\lambda(a, U) \setminus \{T a\}) \subseteq A$
- An “interesting” unfounded set  $U$  satisfies:

$$\emptyset \subset U \subseteq (atom(P) \setminus A^F)$$

- Wrt a fixpoint of unit propagation, such an unfounded set contains some loop of  $P$ 
  - Note Tight programs do not yield “interesting” unfounded sets !
- Given an unfounded set  $U$  and some  $a \in U$ , adding  $\lambda(a, U)$  to  $\nabla$  triggers a conflict or further derivations by unit propagation
  - Note Add loop nogoods atom by atom to eventually falsify all  $a \in U$

# Outline of NogoodPropagation

- Derive deterministic consequences via:
  - Unit propagation on  $\Delta_P$  and  $\nabla$ ;
  - Unfounded sets  $U \subseteq atom(P)$
- Note that  $U$  is **unfounded** if  $EB_P(U) \subseteq A^F$ 
  - Note For any  $a \in U$ , we have  $(\lambda(a, U) \setminus \{T a\}) \subseteq A$
- An “interesting” unfounded set  $U$  satisfies:

$$\emptyset \subset U \subseteq (atom(P) \setminus A^F)$$

- Wrt a fixpoint of unit propagation, such an unfounded set contains some loop of  $P$ 
  - Note Tight programs do not yield “interesting” unfounded sets !
- Given an unfounded set  $U$  and some  $a \in U$ , adding  $\lambda(a, U)$  to  $\nabla$  triggers a conflict or further derivations by unit propagation
  - Note Add loop nogoods atom by atom to eventually falsify all  $a \in U$

# Outline of NogoodPropagation

- Derive deterministic consequences via:
  - Unit propagation on  $\Delta_P$  and  $\nabla$ ;
  - Unfounded sets  $U \subseteq atom(P)$
- Note that  $U$  is **unfounded** if  $EB_P(U) \subseteq A^F$ 
  - Note For any  $a \in U$ , we have  $(\lambda(a, U) \setminus \{T a\}) \subseteq A$
- An “interesting” unfounded set  $U$  satisfies:

$$\emptyset \subset U \subseteq (atom(P) \setminus A^F)$$

- Wrt a fixpoint of unit propagation, such an unfounded set contains some loop of  $P$ 
  - Note Tight programs do not yield “interesting” unfounded sets !
- Given an unfounded set  $U$  and some  $a \in U$ , adding  $\lambda(a, U)$  to  $\nabla$  triggers a conflict or further derivations by unit propagation
  - Note Add loop nogoods atom by atom to eventually falsify all  $a \in U$

---

**Algorithm 3: NOGOODPROPAGATION**


---

**Input** : A normal program  $P$ , a set  $\nabla$  of nogoods, and an assignment  $A$ .

**Output** : An extended assignment and set of nogoods.

$U := \emptyset$  *// unfounded set*

**loop**

**repeat**

**if**  $\delta \subseteq A$  **for some**  $\delta \in \Delta_P \cup \nabla$  **then return**  $(A, \nabla)$  *// conflict*

$\Sigma := \{\delta \in \Delta_P \cup \nabla \mid \delta \setminus A = \{\bar{\sigma}\}, \sigma \notin A\}$  *// unit-resulting nogoods*

**if**  $\Sigma \neq \emptyset$  **then let**  $\bar{\sigma} \in \delta \setminus A$  **for some**  $\delta \in \Sigma$  **in**

$dlevel(\sigma) := \max(\{dlevel(\rho) \mid \rho \in \delta \setminus \{\bar{\sigma}\}\} \cup \{0\})$

$A := A \circ \sigma$

**until**  $\Sigma = \emptyset$

**if**  $loop(P) = \emptyset$  **then return**  $(A, \nabla)$

$U := U \setminus A^F$

**if**  $U = \emptyset$  **then**  $U := \text{UNFOUNDEDSET}(P, A)$

**if**  $U = \emptyset$  **then return**  $(A, \nabla)$  *// no unfounded set  $\emptyset \subset U \subseteq \text{atom}(P) \setminus A^F$*

**let**  $a \in U$  **in**

$\nabla := \nabla \cup \{\{T_a\} \cup \{FB \mid B \in EB_P(U)\}\}$  *// record loop nogood*

---

## Requirements for UNFOUNDEDSET

- Implementations of UNFOUNDEDSET must guarantee the following for a result  $U$ 
  - 1  $U \subseteq (atom(P) \setminus A^F)$
  - 2  $EB_P(U) \subseteq A^F$
  - 3  $U = \emptyset$  iff there is no nonempty unfounded subset of  $(atom(P) \setminus A^F)$
- Beyond that, there are various alternatives, such as:
  - Calculating the greatest unfounded set
  - Calculating unfounded sets within strongly connected components of the positive atom dependency graph of  $P$
  - Usually, the latter option is implemented in ASP solvers

## Requirements for UNFOUNDEDSET

- Implementations of UNFOUNDEDSET must guarantee the following for a result  $U$ 
  - 1  $U \subseteq (\text{atom}(P) \setminus A^F)$
  - 2  $EB_P(U) \subseteq A^F$
  - 3  $U = \emptyset$  iff there is no nonempty unfounded subset of  $(\text{atom}(P) \setminus A^F)$
- Beyond that, there are various alternatives, such as:
  - Calculating the greatest unfounded set
  - Calculating unfounded sets within strongly connected components of the positive atom dependency graph of  $P$
  - Usually, the latter option is implemented in ASP solvers

# Example: NogoodPropagation

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $T\{\sim x\}$ $Ty$ $T\{v\}$ $T\{u, y\}$ $Tv$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, F\{x, y\}\} \in \Delta(\{x, y\})$ $\{F\{\sim x\}, Fx\} = \delta(\{\sim x\})$ $\{F\{\sim y\}, Fy\} = \delta(\{\sim y\})$ $\{Tu, F\{x, y\}, F\{v\}\} = \delta(u)$ $\{F\{u, y\}, Tu, Ty\} = \delta(\{u, y\})$ $\{Fv, T\{u, y\}\} \in \Delta(v)$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

# Outline

- 1 Motivation
- 2 Boolean constraints
- 3 Nogoods from logic programs
- 4 Conflict-driven nogood learning
  - CDNL-ASP Algorithm
  - Nogood Propagation
  - Conflict Analysis



## Outline of Conflict Analysis

- Conflict analysis is triggered whenever some nogood  $\delta \in \Delta_P \cup \nabla$  becomes violated, viz.  $\delta \subseteq A$ , at a decision level  $dl > 0$ 
  - Note that all but the first literal assigned at  $dl$  have been unit-resulting for nogoods  $\varepsilon \in \Delta_P \cup \nabla$
  - If  $\sigma \in \delta$  has been unit-resulting for  $\varepsilon$ , we obtain a new violated nogood by resolving  $\delta$  and  $\varepsilon$  as follows:

$$(\delta \setminus \{\sigma\}) \cup (\varepsilon \setminus \{\bar{\sigma}\})$$

- Resolution is directed by resolving first over the literal  $\sigma \in \delta$  derived last, viz.  $(\delta \setminus A[\sigma]) = \{\sigma\}$ 
  - Iterated resolution progresses in inverse order of assignment
- Iterated resolution stops as soon as it generates a nogood  $\delta$  containing exactly one literal  $\sigma$  assigned at decision level  $dl$ 
  - This literal  $\sigma$  is called First Unique Implication Point (First-UIP)
  - All literals in  $(\delta \setminus \{\sigma\})$  are assigned at decision levels smaller than  $dl$

## Outline of Conflict Analysis

- Conflict analysis is triggered whenever some nogood  $\delta \in \Delta_P \cup \nabla$  becomes violated, viz.  $\delta \subseteq A$ , at a decision level  $dl > 0$ 
  - Note that all but the first literal assigned at  $dl$  have been unit-resulting for nogoods  $\varepsilon \in \Delta_P \cup \nabla$
  - If  $\sigma \in \delta$  has been unit-resulting for  $\varepsilon$ , we obtain a new violated nogood by resolving  $\delta$  and  $\varepsilon$  as follows:

$$(\delta \setminus \{\sigma\}) \cup (\varepsilon \setminus \{\bar{\sigma}\})$$

- Resolution is directed by resolving first over the literal  $\sigma \in \delta$  derived last, viz.  $(\delta \setminus A[\sigma]) = \{\sigma\}$ 
  - Iterated resolution progresses in inverse order of assignment
- Iterated resolution stops as soon as it generates a nogood  $\delta$  containing exactly one literal  $\sigma$  assigned at decision level  $dl$ 
  - This literal  $\sigma$  is called First Unique Implication Point (First-UIP)
  - All literals in  $(\delta \setminus \{\sigma\})$  are assigned at decision levels smaller than  $dl$

## Outline of Conflict Analysis

- Conflict analysis is triggered whenever some nogood  $\delta \in \Delta_P \cup \nabla$  becomes violated, viz.  $\delta \subseteq A$ , at a decision level  $dl > 0$ 
  - Note that all but the first literal assigned at  $dl$  have been unit-resulting for nogoods  $\varepsilon \in \Delta_P \cup \nabla$
  - If  $\sigma \in \delta$  has been unit-resulting for  $\varepsilon$ , we obtain a new violated nogood by resolving  $\delta$  and  $\varepsilon$  as follows:

$$(\delta \setminus \{\sigma\}) \cup (\varepsilon \setminus \{\bar{\sigma}\})$$

- Resolution is directed by resolving first over the literal  $\sigma \in \delta$  derived last, viz.  $(\delta \setminus A[\sigma]) = \{\sigma\}$ 
  - Iterated resolution progresses in inverse order of assignment
- Iterated resolution stops as soon as it generates a nogood  $\delta$  containing exactly one literal  $\sigma$  assigned at decision level  $dl$ 
  - This literal  $\sigma$  is called **First Unique Implication Point** (First-UIP)
  - All literals in  $(\delta \setminus \{\sigma\})$  are assigned at decision levels smaller than  $dl$

---

**Algorithm 4: CONFLICTANALYSIS**


---

**Input** : A non-empty violated nogood  $\delta$ , a normal program  $P$ , a set  $\nabla$  of nogoods, and an assignment  $A$ .

**Output** : A derived nogood and a decision level.

**loop**

**let**  $\sigma \in \delta$  **such that**  $\delta \setminus A[\sigma] = \{\sigma\}$  **in**

$k := \max(\{dlevel(\rho) \mid \rho \in \delta \setminus \{\sigma\}\} \cup \{0\})$

**if**  $k = dlevel(\sigma)$  **then**

**let**  $\varepsilon \in \Delta_P \cup \nabla$  **such that**  $\varepsilon \setminus A[\sigma] = \{\bar{\sigma}\}$  **in**

$\delta := (\delta \setminus \{\sigma\}) \cup (\varepsilon \setminus \{\bar{\sigma}\})$

*// resolution*

**else return**  $(\delta, k)$

---

# Example: ConflictAnalysis

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $T\{\sim x\}$ $Ty$ $T\{v\}$ $T\{u, y\}$ $Tv$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\{F\{\sim x\}, Fx\} = \delta(\{\sim x\})$ $\{F\{\sim y\}, Fy\} = \delta(\{\sim y\})$ $\{Tu, F\{x, y\}, F\{v\}\} = \delta(u)$ $\{F\{u, y\}, Tu, Ty\} = \delta(\{u, y\})$ $\{Fv, T\{u, y\}\} \in \Delta(v)$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

$$\{Tu, Fx\}$$

$$\{Tu, Fx, F\{x\}\}$$

✗

# Example: ConflictAnalysis

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $T\{\sim x\}$ $Ty$ $T\{v\}$ $T\{u, y\}$ $Tv$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\{F\{\sim x\}, Fx\} = \delta(\{\sim x\})$ $\{F\{\sim y\}, Fy\} = \delta(\{\sim y\})$ $\{Tu, F\{x, y\}, F\{v\}\} = \delta(u)$ $\{F\{u, y\}, Tu, Ty\} = \delta(\{u, y\})$ $\{Fv, T\{u, y\}\} \in \Delta(v)$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

$$\{Tu, Fx\}$$

$$\{Tu, Fx, F\{x\}\}$$

✗

# Example: ConflictAnalysis

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $T\{\sim x\}$ $Ty$ $T\{v\}$ $T\{u, y\}$ $Tv$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\{F\{\sim x\}, Fx\} = \delta(\{\sim x\})$ $\{F\{\sim y\}, Fy\} = \delta(\{\sim y\})$ $\{Tu, F\{x, y\}, F\{v\}\} = \delta(u)$ $\{F\{u, y\}, Tu, Ty\} = \delta(\{u, y\})$ $\{Fv, T\{u, y\}\} \in \Delta(v)$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

$$\{Tu, Fx\}$$

$$\{Tu, Fx, F\{x\}\}$$

**X**

# Example: ConflictAnalysis

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $T\{\sim x\}$ $Ty$ $T\{v\}$ $T\{u, y\}$ $Tv$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\{F\{\sim x\}, Fx\} = \delta(\{\sim x\})$ $\{F\{\sim y\}, Fy\} = \delta(\{\sim y\})$ $\{Tu, F\{x, y\}, F\{v\}\} = \delta(u)$ $\{F\{u, y\}, Tu, Ty\} = \delta(\{u, y\})$ $\{Fv, T\{u, y\}\} \in \Delta(v)$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

$$\{Tu, Fx\}$$

$$\{Tu, Fx, F\{x\}\}$$

✗



# Example: ConflictAnalysis

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $T\{\sim x\}$ $Ty$ $T\{v\}$ $T\{u, y\}$ $Tv$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\{F\{\sim x\}, Fx\} = \delta(\{\sim x\})$ $\{F\{\sim y\}, Fy\} = \delta(\{\sim y\})$ $\{Tu, F\{x, y\}, F\{v\}\} = \delta(u)$ $\{F\{u, y\}, Tu, Ty\} = \delta(\{u, y\})$ $\{Fv, T\{u, y\}\} \in \Delta(v)$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

$$\{Tu, Fx\}$$

$$\{Tu, Fx, F\{x\}\}$$

✗

# Example: ConflictAnalysis

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $T\{\sim x\}$ $Ty$ $T\{v\}$ $T\{u, y\}$ $Tv$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\{F\{\sim x\}, Fx\} = \delta(\{\sim x\})$ $\{F\{\sim y\}, Fy\} = \delta(\{\sim y\})$ $\{Tu, F\{x, y\}, F\{v\}\} = \delta(u)$ $\{F\{u, y\}, Tu, Ty\} = \delta(\{u, y\})$ $\{Fv, T\{u, y\}\} \in \Delta(v)$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

$$\{Tu, Fx\}$$

$$\{Tu, Fx, F\{x\}\}$$

✗

# Example: ConflictAnalysis

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $T\{\sim x\}$ $Ty$ $T\{v\}$ $T\{u, y\}$ $Tv$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\{F\{\sim x\}, Fx\} = \delta(\{\sim x\})$ $\{F\{\sim y\}, Fy\} = \delta(\{\sim y\})$ $\{Tu, F\{x, y\}, F\{v\}\} = \delta(u)$ $\{F\{u, y\}, Tu, Ty\} = \delta(\{u, y\})$ $\{Fv, T\{u, y\}\} \in \Delta(v)$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

$$\{Tu, Fx\}$$

$$\{Tu, Fx, F\{x\}\}$$

✗

# Example: ConflictAnalysis

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	<b><math>Tu</math></b>		
2	<b><math>F\{\sim x, \sim y\}</math></b>	<b><math>Fw</math></b>	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	<b><math>F\{\sim y\}</math></b>	<b><math>Fx</math></b> <b><math>F\{x\}</math></b> <b><math>F\{x, y\}</math></b> <b><math>T\{\sim x\}</math></b> <b><math>Ty</math></b> <b><math>T\{v\}</math></b> <b><math>T\{u, y\}</math></b> <b><math>Tv</math></b>	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\{F\{\sim x\}, Fx\} = \delta(\{\sim x\})$ $\{F\{\sim y\}, Fy\} = \delta(\{\sim y\})$ $\{Tu, F\{x, y\}, F\{v\}\} = \delta(u)$ $\{F\{u, y\}, Tu, Ty\} = \delta(\{u, y\})$ $\{Fv, T\{u, y\}\} \in \Delta(v)$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

$$\{Tu, Fx\}$$

$$\{Tu, Fx, F\{x\}\}$$

**x**

# Example: ConflictAnalysis

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $T\{\sim x\}$ $Ty$ $T\{v\}$ $T\{u, y\}$ $Tv$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\{F\{\sim x\}, Fx\} = \delta(\{\sim x\})$ $\{F\{\sim y\}, Fy\} = \delta(\{\sim y\})$ $\{Tu, F\{x, y\}, F\{v\}\} = \delta(u)$ $\{F\{u, y\}, Tu, Ty\} = \delta(\{u, y\})$ $\{Fv, T\{u, y\}\} \in \Delta(v)$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

$\{Tu, Fx\}$   
 $\{Tu, Fx, F\{x\}\}$

**x**

# Example: ConflictAnalysis

Consider

$$P = \left\{ \begin{array}{llll} x \leftarrow \sim y & u \leftarrow x, y & v \leftarrow x & w \leftarrow \sim x, \sim y \\ y \leftarrow \sim x & u \leftarrow v & v \leftarrow u, y & \end{array} \right\}$$

$dl$	$\sigma_d$	$\bar{\sigma}$	$\delta$
1	$Tu$		
2	$F\{\sim x, \sim y\}$	$Fw$	$\{Tw, F\{\sim x, \sim y\}\} = \delta(w)$
3	$F\{\sim y\}$	$Fx$ $F\{x\}$ $F\{x, y\}$ $T\{\sim x\}$ $Ty$ $T\{v\}$ $T\{u, y\}$ $Tv$	$\{Tx, F\{\sim y\}\} = \delta(x)$ $\{T\{x\}, Fx\} \in \Delta(\{x\})$ $\{T\{x, y\}, Fx\} \in \Delta(\{x, y\})$ $\{F\{\sim x\}, Fx\} = \delta(\{\sim x\})$ $\{F\{\sim y\}, Fy\} = \delta(\{\sim y\})$ $\{Tu, F\{x, y\}, F\{v\}\} = \delta(u)$ $\{F\{u, y\}, Tu, Ty\} = \delta(\{u, y\})$ $\{Fv, T\{u, y\}\} \in \Delta(v)$ $\{Tu, F\{x\}, F\{x, y\}\} = \lambda(u, \{u, v\})$

$$\{Tu, Fx\}$$

$$\{Tu, Fx, F\{x\}\}$$

**x**

## Remarks

- There always is a First-UIP at which conflict analysis terminates
  - In the worst, resolution stops at the heuristically chosen literal assigned at decision level  $dl$
- The nogood  $\delta$  containing First-UIP  $\sigma$  is violated by  $A$ , viz.  $\delta \subseteq A$
- We have  $k = \max(\{dl(\rho) \mid \rho \in \delta \setminus \{\sigma\}\} \cup \{0\}) < dl$ 
  - After recording  $\delta$  in  $\nabla$  and backjumping to decision level  $k$ ,  $\bar{\sigma}$  is unit-resulting for  $\delta$  !
  - Such a nogood  $\delta$  is called asserting
- Asserting nogoods direct conflict-driven search into a different region of the search space than traversed before, without explicitly flipping any heuristically chosen literal !

## Remarks

- There always is a First-UIP at which conflict analysis terminates
  - In the worst, resolution stops at the heuristically chosen literal assigned at decision level  $dl$
- The nogood  $\delta$  containing First-UIP  $\sigma$  is violated by  $A$ , viz.  $\delta \subseteq A$
- We have  $k = \max(\{dl(\rho) \mid \rho \in \delta \setminus \{\sigma\}\} \cup \{0\}) < dl$ 
  - After recording  $\delta$  in  $\nabla$  and backjumping to decision level  $k$ ,  $\bar{\sigma}$  is unit-resulting for  $\delta$  !
  - Such a nogood  $\delta$  is called asserting
- Asserting nogoods direct conflict-driven search into a different region of the search space than traversed before, without explicitly flipping any heuristically chosen literal !



## Remarks

- There always is a First-UIP at which conflict analysis terminates
  - In the worst, resolution stops at the heuristically chosen literal assigned at decision level  $dl$
- The nogood  $\delta$  containing First-UIP  $\sigma$  is violated by  $A$ , viz.  $\delta \subseteq A$
- We have  $k = \max(\{dl(\rho) \mid \rho \in \delta \setminus \{\sigma\}\} \cup \{0\}) < dl$ 
  - After recording  $\delta$  in  $\nabla$  and backjumping to decision level  $k$ ,  $\bar{\sigma}$  is unit-resulting for  $\delta$  !
  - Such a nogood  $\delta$  is called **asserting**
- Asserting nogoods direct conflict-driven search into a different region of the search space than traversed before, without explicitly flipping any heuristically chosen literal !

## Remarks

- There always is a First-UIP at which conflict analysis terminates
  - In the worst, resolution stops at the heuristically chosen literal assigned at decision level  $dl$
- The nogood  $\delta$  containing First-UIP  $\sigma$  is violated by  $A$ , viz.  $\delta \subseteq A$
- We have  $k = \max(\{dl(\rho) \mid \rho \in \delta \setminus \{\sigma\}\} \cup \{0\}) < dl$ 
  - After recording  $\delta$  in  $\nabla$  and backjumping to decision level  $k$ ,  $\bar{\sigma}$  is unit-resulting for  $\delta$  !
  - Such a nogood  $\delta$  is called **asserting**
- Asserting nogoods direct conflict-driven search into a different region of the search space than traversed before, without explicitly flipping any heuristically chosen literal !

- [1] C. Anger, M. Gebser, T. Linke, A. Neumann, and T. Schaub.  
**The nomore++ approach to answer set solving.**  
In G. Sutcliffe and A. Voronkov, editors, *Proceedings of the Twelfth International Conference on Logic for Programming, Artificial Intelligence, and Reasoning (LPAR'05)*, volume 3835 of *Lecture Notes in Artificial Intelligence*, pages 95–109. Springer-Verlag, 2005.
- [2] C. Anger, K. Konczak, T. Linke, and T. Schaub.  
**A glimpse of answer set programming.**  
*Künstliche Intelligenz*, 19(1):12–17, 2005.
- [3] Y. Babovich and V. Lifschitz.  
**Computing answer sets using program completion.**  
Unpublished draft, 2003.
- [4] C. Baral.  
***Knowledge Representation, Reasoning and Declarative Problem Solving.***  
Cambridge University Press, 2003.

- [5] C. Baral, G. Brewka, and J. Schlipf, editors.  
*Proceedings of the Ninth International Conference on Logic Programming and Nonmonotonic Reasoning (LPNMR'07)*, volume 4483 of *Lecture Notes in Artificial Intelligence*. Springer-Verlag, 2007.
- [6] C. Baral and M. Gelfond.  
Logic programming and knowledge representation.  
*Journal of Logic Programming*, 12:1–80, 1994.
- [7] S. Baselice, P. Bonatti, and M. Gelfond.  
Towards an integration of answer set and constraint solving.  
In M. Gabbrielli and G. Gupta, editors, *Proceedings of the Twenty-first International Conference on Logic Programming (ICLP'05)*, volume 3668 of *Lecture Notes in Computer Science*, pages 52–66. Springer-Verlag, 2005.
- [8] A. Biere.  
Adaptive restart strategies for conflict driven SAT solvers.

In H. Kleine Büning and X. Zhao, editors, *Proceedings of the Eleventh International Conference on Theory and Applications of Satisfiability Testing (SAT'08)*, volume 4996 of *Lecture Notes in Computer Science*, pages 28–33. Springer-Verlag, 2008.

- [9] A. Biere.  
**PicoSAT essentials.**  
*Journal on Satisfiability, Boolean Modeling and Computation*, 4:75–97, 2008.
- [10] A. Biere, M. Heule, H. van Maaren, and T. Walsh, editors.  
**Handbook of Satisfiability**, volume 185 of *Frontiers in Artificial Intelligence and Applications*.  
IOS Press, 2009.
- [11] G. Brewka, T. Eiter, and M. Truszczynski.  
**Answer set programming at a glance.**  
*Communications of the ACM*, 54(12):92–103, 2011.
- [12] K. Clark.

## Negation as failure.

In H. Gallaire and J. Minker, editors, *Logic and Data Bases*, pages 293–322. Plenum Press, 1978.

- [13] M. D'Agostino, D. Gabbay, R. Hähnle, and J. Posegga, editors. *Handbook of Tableau Methods*. Kluwer Academic Publishers, 1999.
- [14] E. Dantsin, T. Eiter, G. Gottlob, and A. Voronkov. **Complexity and expressive power of logic programming.** In *Proceedings of the Twelfth Annual IEEE Conference on Computational Complexity (CCC'97)*, pages 82–101. IEEE Computer Society Press, 1997.
- [15] M. Davis, G. Logemann, and D. Loveland. **A machine program for theorem-proving.** *Communications of the ACM*, 5:394–397, 1962.
- [16] M. Davis and H. Putnam. **A computing procedure for quantification theory.**

*Journal of the ACM*, 7:201–215, 1960.

- [17] C. Drescher, M. Gebser, T. Grote, B. Kaufmann, A. König, M. Ostrowski, and T. Schaub.

**Conflict-driven disjunctive answer set solving.**

In G. Brewka and J. Lang, editors, *Proceedings of the Eleventh International Conference on Principles of Knowledge Representation and Reasoning (KR'08)*, pages 422–432. AAAI Press, 2008.

- [18] C. Drescher, M. Gebser, B. Kaufmann, and T. Schaub.

**Heuristics in conflict resolution.**

In M. Pagnucco and M. Thielscher, editors, *Proceedings of the Twelfth International Workshop on Nonmonotonic Reasoning (NMR'08)*, number UNSW-CSE-TR-0819 in School of Computer Science and Engineering, The University of New South Wales, Technical Report Series, pages 141–149, 2008.

- [19] N. Eén and N. Sörensson.

**An extensible SAT-solver.**

In E. Giunchiglia and A. Tacchella, editors, *Proceedings of the Sixth International Conference on Theory and Applications of Satisfiability Testing (SAT'03)*, volume 2919 of *Lecture Notes in Computer Science*, pages 502–518. Springer-Verlag, 2004.

[20] T. Eiter and G. Gottlob.

**On the computational cost of disjunctive logic programming:  
Propositional case.**

*Annals of Mathematics and Artificial Intelligence*, 15(3-4):289–323, 1995.

[21] T. Eiter, G. Ianni, and T. Krennwallner.

**Answer Set Programming: A Primer.**

In S. Tessaris, E. Franconi, T. Eiter, C. Gutierrez, S. Handschuh, M. Rousset, and R. Schmidt, editors, *Fifth International Reasoning Web Summer School (RW'09)*, volume 5689 of *Lecture Notes in Computer Science*, pages 40–110. Springer-Verlag, 2009.

[22] F. Fages.

**Consistency of Clark's completion and the existence of stable models.**



*Journal of Methods of Logic in Computer Science*, 1:51–60, 1994.

[23] P. Ferraris.

**Answer sets for propositional theories.**

In C. Baral, G. Greco, N. Leone, and G. Terracina, editors, *Proceedings of the Eighth International Conference on Logic Programming and Nonmonotonic Reasoning (LPNMR'05)*, volume 3662 of *Lecture Notes in Artificial Intelligence*, pages 119–131. Springer-Verlag, 2005.

[24] P. Ferraris and V. Lifschitz.

**Mathematical foundations of answer set programming.**

In S. Artëmov, H. Barringer, A. d'Avila Garcez, L. Lamb, and J. Woods, editors, *We Will Show Them! Essays in Honour of Dov Gabbay*, volume 1, pages 615–664. College Publications, 2005.

[25] M. Fitting.

**A Kripke-Kleene semantics for logic programs.**

*Journal of Logic Programming*, 2(4):295–312, 1985.

- [26] M. Gebser, R. Kaminski, B. Kaufmann, M. Ostrowski, T. Schaub, and S. Thiele.

*A user's guide to gringo, clasp, clingo, and iclingo.*

- [27] M. Gebser, R. Kaminski, B. Kaufmann, M. Ostrowski, T. Schaub, and S. Thiele.

*Engineering an incremental ASP solver.*

In M. Garcia de la Banda and E. Pontelli, editors, *Proceedings of the Twenty-fourth International Conference on Logic Programming (ICLP'08)*, volume 5366 of *Lecture Notes in Computer Science*, pages 190–205. Springer-Verlag, 2008.

- [28] M. Gebser, R. Kaminski, B. Kaufmann, and T. Schaub.

*On the implementation of weight constraint rules in conflict-driven ASP solvers.*

In Hill and Warren [44], pages 250–264.

- [29] M. Gebser, R. Kaminski, B. Kaufmann, and T. Schaub.

*Answer Set Solving in Practice.*

Synthesis Lectures on Artificial Intelligence and Machine Learning.  
Morgan and Claypool Publishers, 2012.

- [30] M. Gebser, B. Kaufmann, A. Neumann, and T. Schaub.  
**clasp: A conflict-driven answer set solver.**  
In Baral et al. [5], pages 260–265.
- [31] M. Gebser, B. Kaufmann, A. Neumann, and T. Schaub.  
**Conflict-driven answer set enumeration.**  
In Baral et al. [5], pages 136–148.
- [32] M. Gebser, B. Kaufmann, A. Neumann, and T. Schaub.  
**Conflict-driven answer set solving.**  
In Veloso [68], pages 386–392.
- [33] M. Gebser, B. Kaufmann, A. Neumann, and T. Schaub.  
**Advanced preprocessing for answer set solving.**  
In M. Ghallab, C. Spyropoulos, N. Fakotakis, and N. Avouris, editors,  
*Proceedings of the Eighteenth European Conference on Artificial  
Intelligence (ECAI'08)*, pages 15–19. IOS Press, 2008.

- [34] M. Gebser, B. Kaufmann, and T. Schaub.  
**The conflict-driven answer set solver clasp: Progress report.**  
In E. Erdem, F. Lin, and T. Schaub, editors, *Proceedings of the Tenth International Conference on Logic Programming and Nonmonotonic Reasoning (LPNMR'09)*, volume 5753 of *Lecture Notes in Artificial Intelligence*, pages 509–514. Springer-Verlag, 2009.
- [35] M. Gebser, B. Kaufmann, and T. Schaub.  
**Solution enumeration for projected Boolean search problems.**  
In W. van Hoes and J. Hooker, editors, *Proceedings of the Sixth International Conference on Integration of AI and OR Techniques in Constraint Programming for Combinatorial Optimization Problems (CPAIOR'09)*, volume 5547 of *Lecture Notes in Computer Science*, pages 71–86. Springer-Verlag, 2009.
- [36] M. Gebser, M. Ostrowski, and T. Schaub.  
**Constraint answer set solving.**  
In Hill and Warren [44], pages 235–249.

[37] M. Gebser and T. Schaub.

**Tableau calculi for answer set programming.**

In S. Etalle and M. Truszczynski, editors, *Proceedings of the Twenty-second International Conference on Logic Programming (ICLP'06)*, volume 4079 of *Lecture Notes in Computer Science*, pages 11–25. Springer-Verlag, 2006.

[38] M. Gebser and T. Schaub.

**Generic tableaux for answer set programming.**

In V. Dahl and I. Niemelä, editors, *Proceedings of the Twenty-third International Conference on Logic Programming (ICLP'07)*, volume 4670 of *Lecture Notes in Computer Science*, pages 119–133. Springer-Verlag, 2007.

[39] M. Gelfond.

**Answer sets.**

In V. Lifschitz, F. van Harmelen, and B. Porter, editors, *Handbook of Knowledge Representation*, chapter 7, pages 285–316. Elsevier Science, 2008.

- [40] M. Gelfond and N. Leone.  
Logic programming and knowledge representation — the A-Prolog perspective.  
*Artificial Intelligence*, 138(1-2):3–38, 2002.
- [41] M. Gelfond and V. Lifschitz.  
The stable model semantics for logic programming.  
In R. Kowalski and K. Bowen, editors, *Proceedings of the Fifth International Conference and Symposium of Logic Programming (ICLP'88)*, pages 1070–1080. MIT Press, 1988.
- [42] M. Gelfond and V. Lifschitz.  
Logic programs with classical negation.  
In D. Warren and P. Szeredi, editors, *Proceedings of the Seventh International Conference on Logic Programming (ICLP'90)*, pages 579–597. MIT Press, 1990.
- [43] E. Giunchiglia, Y. Lierler, and M. Maratea.  
Answer set programming based on propositional satisfiability.

*Journal of Automated Reasoning*, 36(4):345–377, 2006.

- [44] P. Hill and D. Warren, editors.  
*Proceedings of the Twenty-fifth International Conference on Logic Programming (ICLP'09)*, volume 5649 of *Lecture Notes in Computer Science*. Springer-Verlag, 2009.
- [45] J. Huang.  
The effect of restarts on the efficiency of clause learning.  
In Veloso [68], pages 2318–2323.
- [46] K. Konczak, T. Linke, and T. Schaub.  
Graphs and colorings for answer set programming.  
*Theory and Practice of Logic Programming*, 6(1-2):61–106, 2006.
- [47] J. Lee.  
A model-theoretic counterpart of loop formulas.  
In L. Kaelbling and A. Saffiotti, editors, *Proceedings of the Nineteenth International Joint Conference on Artificial Intelligence (IJCAI'05)*, pages 503–508. Professional Book Center, 2005.

- [48] N. Leone, G. Pfeifer, W. Faber, T. Eiter, G. Gottlob, S. Perri, and F. Scarcello.

**The DLV system for knowledge representation and reasoning.**

*ACM Transactions on Computational Logic*, 7(3):499–562, 2006.

- [49] V. Lifschitz.

**Answer set programming and plan generation.**

*Artificial Intelligence*, 138(1-2):39–54, 2002.

- [50] V. Lifschitz.

**Introduction to answer set programming.**

Unpublished draft, 2004.

- [51] V. Lifschitz and A. Razborov.

**Why are there so many loop formulas?**

*ACM Transactions on Computational Logic*, 7(2):261–268, 2006.

- [52] F. Lin and Y. Zhao.

**ASSAT: computing answer sets of a logic program by SAT solvers.**

*Artificial Intelligence*, 157(1-2):115–137, 2004.



- [53] V. Marek and M. Truszczyński.  
*Nonmonotonic logic: context-dependent reasoning.*  
Artificial Intelligence. Springer-Verlag, 1993.
- [54] V. Marek and M. Truszczyński.  
**Stable models and an alternative logic programming paradigm.**  
In K. Apt, V. Marek, M. Truszczyński, and D. Warren, editors, *The Logic Programming Paradigm: a 25-Year Perspective*, pages 375–398.  
Springer-Verlag, 1999.
- [55] J. Marques-Silva, I. Lynce, and S. Malik.  
**Conflict-driven clause learning SAT solvers.**  
In Biere et al. [10], chapter 4, pages 131–153.
- [56] J. Marques-Silva and K. Sakallah.  
**GRASP: A search algorithm for propositional satisfiability.**  
*IEEE Transactions on Computers*, 48(5):506–521, 1999.
- [57] V. Mellarkod and M. Gelfond.  
**Integrating answer set reasoning with constraint solving techniques.**

In J. Garrigue and M. Hermenegildo, editors, *Proceedings of the Ninth International Symposium on Functional and Logic Programming (FLOPS'08)*, volume 4989 of *Lecture Notes in Computer Science*, pages 15–31. Springer-Verlag, 2008.

- [58] V. Mellarkod, M. Gelfond, and Y. Zhang.  
**Integrating answer set programming and constraint logic programming.**  
*Annals of Mathematics and Artificial Intelligence*, 53(1-4):251–287, 2008.
- [59] D. Mitchell.  
**A SAT solver primer.**  
*Bulletin of the European Association for Theoretical Computer Science*, 85:112–133, 2005.
- [60] M. Moskewicz, C. Madigan, Y. Zhao, L. Zhang, and S. Malik.  
**Chaff: Engineering an efficient SAT solver.**  
In *Proceedings of the Thirty-eighth Conference on Design Automation (DAC'01)*, pages 530–535. ACM Press, 2001.

- [61] I. Niemelä.  
Logic programs with stable model semantics as a constraint programming paradigm.  
*Annals of Mathematics and Artificial Intelligence*, 25(3-4):241–273, 1999.
- [62] R. Nieuwenhuis, A. Oliveras, and C. Tinelli.  
Solving SAT and SAT modulo theories: From an abstract Davis-Putnam-Logemann-Loveland procedure to DPLL(T).  
*Journal of the ACM*, 53(6):937–977, 2006.
- [63] K. Pipatsrisawat and A. Darwiche.  
A lightweight component caching scheme for satisfiability solvers.  
In J. Marques-Silva and K. Sakallah, editors, *Proceedings of the Tenth International Conference on Theory and Applications of Satisfiability Testing (SAT'07)*, volume 4501 of *Lecture Notes in Computer Science*, pages 294–299. Springer-Verlag, 2007.
- [64] L. Ryan.

## Efficient algorithms for clause-learning SAT solvers.

Master's thesis, Simon Fraser University, 2004.

- [65] P. Simons, I. Niemelä, and T. Soinen.  
Extending and implementing the stable model semantics.  
*Artificial Intelligence*, 138(1-2):181–234, 2002.
- [66] T. Syrjänen.  
Lparse 1.0 user's manual.
- [67] A. Van Gelder, K. Ross, and J. Schlipf.  
The well-founded semantics for general logic programs.  
*Journal of the ACM*, 38(3):620–650, 1991.
- [68] M. Veloso, editor.  
*Proceedings of the Twentieth International Joint Conference on Artificial Intelligence (IJCAI'07)*. AAAI/MIT Press, 2007.
- [69] L. Zhang, C. Madigan, M. Moskewicz, and S. Malik.  
Efficient conflict driven learning in a Boolean satisfiability solver.

In *Proceedings of the International Conference on Computer-Aided Design (ICCAD'01)*, pages 279–285. ACM Press, 2001.