

# Managing Inconsistent Ontologies

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# Outline

- 1 Introduction
- 2 Propositional Knowledge Integration
- 3 Knowledge Integration for DLs ( $\mathcal{ALC}$ )
- 4 Implementation
- 5 Example
- 6 Future Work

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- 1 **Introduction**
- 2 Propositional Knowledge Integration
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# Reasoning with Ontologies

## Ontologies

- Description of concepts and relationships between them
- Description Logics (DLs) are used to represent ontologies
- DLs provide the basis for reasoning with ontologies

## Inconsistency management

- DL reasoners are surprisingly efficient at detecting inconsistencies
- Little guidance for resolving inconsistencies

# Inconsistency Management for DLs

## Propositional knowledge integration

- There is a vast body of literature on knowledge integration
- Applicable to a whole range of logics
- The solutions are typically propositional in nature

## Application to DLs

- Application of existing propositional techniques to DLs
- Exploit the additional structure of DLs to refine these techniques
- Applicable to both inconsistency and concept unsatisfiability

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# Benferhat et al. 2004

## Stratified knowledge bases

- $(S_1, \dots, S_n)$  with each  $S_i$  a finite multi-set of sentences
- Lower indices represent higher reliability

## Knowledge integration based on two principles

- Independence of sentences in the same stratum
- More reliable sentences should take precedence

## Algorithms

- Benferhat et al: Disjunctive Maxi-adjustment
- Our version: Conjunctive Maxi-adjustment (CMA)

# The CMA Algorithm

**Input:**  $K = (S_1, \dots, S_n)$

**Output:** A consistent classical knowledge base

$B := \emptyset$

FOR  $i := 1$  to  $n$

$j := |S_i|$

  REPEAT

$\phi := \bigvee$  of all  $\wedge$ s of size  $j$  of formulas of  $S_i$

$j := j - 1$

  UNTIL  $B \cup \{\phi\}$  is consistent or  $j = 0$

  IF  $B \cup \{\phi\}$  is consistent

$B := B \cup \{\phi\}$

  ENDIF

ENDFOR

return  $B$

# An example of CMA

- 1  $K = (S_1, S_2)$
- 2  $S_1 = \{\neg(p \wedge q), \neg(q \wedge r), \neg(p \wedge r)\}$ ,  $S_2 = \{p, q, r\}$
- 3  $S_1$  is consistent:  $B$  is set to  $S_1$
- 4  $S_2$  is inconsistent with  $B$ :  
 $\phi$  is set to  $(p \wedge q) \vee (p \wedge r) \vee (q \wedge r)$ .
- 5  $\phi$  is inconsistent with  $B$ :  $\phi$  is weakened to  $p \vee q \vee r$
- 6  $\phi$  is consistent with  $B$ :  
 $B$  is set to  $\{\neg(p \wedge q), \neg(q \wedge r), \neg(p \wedge r), p \vee q \vee r\}$

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# Disjunctive DL knowledge bases

## Port CMA to DLs

- DLs are not expressive enough for a direct port
- E.g.  $(C \sqsubseteq D) \sqcup C(a)$  is ill-formed
- This is not an issue for testing concept satisfiability
- Regardless: wish to retain the structure

## Disjunctive DL knowledge bases

- Set of DL knowledge bases  $\mathcal{B}$
- E.g.  $\mathcal{B} = \{[C \sqsubseteq D], [C(a)]\}$
- Meaning:  $\mathcal{B}$  is true iff at least one of its elements are true

## The CMA algorithm for DLs

**Input:**  $K = (S_1, \dots, S_n)$ ; **Output:** A consistent DKB

$\mathcal{B} := \{\emptyset\}$

FOR  $i := 1$  to  $n$

$\mathcal{C} := \mathcal{B}$

    FORALL  $B \in \mathcal{C}$

$j := |S_i|$

        REPEAT

$\mathcal{X} := \{X \mid X \subseteq S_i \text{ and } |X| = j\}$

$j := j - 1$

        UNTIL  $B \cup X$  is consistent for some  $X \in \mathcal{X}$

$\mathcal{B} := (\mathcal{B} \setminus \{B\}) \cup$

$\{B \cup X \mid (X \in \mathcal{X}) \ \& \ (B \cup X) \text{ is consistent}\}$

    ENDFOR

ENDFOR

**return**  $\mathcal{B}$

# CMA for DLS

## An example of CMA for DLs

- 1  $K = (S_1, S_2)$
- 2  $S_1 = [taxPR(Tommie), \neg PR(Tommie), \neg PR(Louise)]$   
 $S_2 = [\neg PR \sqsubseteq \neg taxPR]$
- 3  $S_1$  is consistent:  $\mathcal{B}$  is set to  $\{S_1\}$
- 4  $S_2$  is inconsistent with  $S_1$ , the only element of  $\mathcal{B}$
- 5  $\mathcal{B} = \{[\neg PR(Tommie), taxPR(Tommie), \neg PR(Louise)]\}$

## CMA for DLs does not exploit DL structure

- Ought to conclude that Louise does not have taxPR
- Need to express: everyone without PR, except Tommie, do not have taxPR

# Refined version of CMA for DLs

## Enrich the language

- Replace subsumption with exception-based subsumption
- $C \sqsubseteq_n D$  means the subsumption may have  $n$  exceptions
- $C \sqsubseteq D$  expressed as  $C \sqsubseteq_0 D$
- Can be expressed using nominals:  

$$C \sqsubseteq_n D \equiv ((\leq n) C \sqcap \neg D) \equiv (C \sqcap \neg D) \sqsubseteq \{a_1, \dots, a_n\}$$

## Additional structure allows for weakening statements

- $C \sqsubseteq D$  weakened to  $C \sqsubseteq_1 D$
- In general:  $C \sqsubseteq_n D$  weakened to  $C \sqsubseteq_{n+1} D$

# The refined CMA algorithm for DLs

**Input:**  $K = (S_1, \dots, S_n)$ ; **Output:** A consistent DKB

$\mathcal{B} := \{\emptyset\}$

FOR  $i := 1$  to  $n$

$\mathcal{C} := \mathcal{B}$

    FORALL  $B \in \mathcal{C}$

$j := 0$

        REPEAT

$\mathcal{X} := \text{appropriate weakening}$

$j := j + 1$

        UNTIL  $B \cup X$  is consistent for some  $X \in \mathcal{X}$

$\mathcal{B} := (\mathcal{B} \setminus \{B\}) \cup$

$\{B \cup X \mid (X \in \mathcal{X}) \ \& \ (B \cup X) \text{ is consistent}\}$

    ENDFOR

ENDFOR

return  $\mathcal{B}$

## Example of refined CMA for DLs

- 1  $S_1 = [\neg PR(T), \neg PR(L)]$ , and  
 $S_2 = [TPR(T), TPR(L), \neg PR \sqsubseteq_0 \neg TPR]$
- 2  $S_1$  is consistent:  $\mathcal{B}$  (and  $\mathcal{C}$ ) =  $\{S_1\}$
- 3  $B$  = only element of  $\mathcal{C}$ :  $[\neg PR(T), \neg PR(L)]$
- 4  $S_2$  is inconsistent with  $B$ :  $\mathcal{X} =$   
 $\{[TPR(T), TPR(L), \neg PR \sqsubseteq_1 \neg TPR]$   
 $[TPR(T), \neg PR \sqsubseteq_0 \neg TPR]$   
 $[TPR(L), \neg PR \sqsubseteq_0 \neg TPR]\}$
- 5 Every element of  $\mathcal{X}$  is inconsistent with  $B$ :  $\mathcal{X}$  is set to:  
 $\{[TPR(T), TPR(L), \neg PR \sqsubseteq_2 \neg TPR],$   
 $[TPR(T), \neg PR \sqsubseteq_1 \neg TPR],$   
 $[TPR(L), \neg PR \sqsubseteq_1 \neg TPR],$   
 $[\neg PR \sqsubseteq_0 \neg TPR]\}$

## Example of refined CMA for DLs

### Consistency is reached and exactly one of four cases hold

- 1 Tommie and Louise are the only two people without PR, but with taxPR
- 2 Tommie is the only person without PR, but with taxPR
- 3 Louise is the only person without PR, but with taxPR
- 4 All persons without PR, including Tommie and Louise, do not have taxPR

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# Maximally concept-satisfiable subsets

- Related to finding consistent subsets of maximal cardinality
- Closely related to finding maximally consistent subsets
- Current implementation: maximally concept-satisfiable subsets
- Specialised tableau algorithm for acyclic Tboxes
- Outperforms optimised reasoners using brute force
- Similar principles apply for pinpointing

# The Specialised Tableau Algorithm

## Algorithm for finding MSSs (relative to $A_j$ )

- $\Gamma$  has  $n$  axioms of the form  $A_i \doteq C_i$
- The algorithm generates a tree in classical tableau style
- Nodes are labelled with indexed concept assertions
- $(a : C, I)$ : axioms indexed by  $I$  are responsible for  $a : C$
- Root node contains  $(a : A_j, \emptyset)$
- The tree is generated from a number of expansion rules
- Each node  $x$  maintains an exclusion-set  $E(x)$
- Performs conversion to NNF on the fly

# Expansion Rules

## 1. $D^+$ -rule

- If  $(a : A_i, I)$  is in  $L(x)$  and has not been tagged then
  - Tag  $(a : A_i, I)$  and  $L(x) := L(x) \cup \{(a : C_i, I \cup \{i\})\}$
- Lazy unfolding for positive concepts
- Observe the updating of the index set

## 2. $D^-$ -rule

- If  $(a : \neg A_i, I)$  is in  $L(x)$  and has not been tagged then
  - Tag  $(a : \neg A_i, I)$  and  $L(x) := L(x) \cup \{(a : \neg C_i, I \cup \{i\})\}$
- Lazy unfolding for negative concepts
- Observe the updating of the index set

# Expansion Rules

## 3. $\sqcap$ -rule

- If  $(a : C \sqcap D, I) \in L(x)$  then
  - $L(x) := L(x) \setminus \{(a : C \sqcap D, I)\} \cup \{(a : C, I), (a : D, I)\}$
- Classical  $\sqcap$ -rule

## 4. $\sqcup$ -rule

- If  $(a : C \sqcup D, I) \in L(x)$  then
  - Create two children  $y$  and  $z$  of  $x$
  - $L(y) := L(x) \setminus \{(a : C \sqcup D, I)\} \cup \{(a : C, I)\}; E(y) := E(x)$
  - $L(z) := L(x) \setminus \{(a : C \sqcup D, I)\} \cup \{(a : D, I)\}; E(z) := E(x)$
- Classical  $\sqcup$ -rule

# Expansion Rules

## 5. $\exists$ -rule

- If  $(a : \exists R.C, I) \in L(x)$  and rules 1-4 can't be applied then
  - $X := \{(b : C, I)\} \cup \{(b : D, I \cup J) \mid (a : \forall R.D, J) \in L(x)\}$
  - ( $b$  is a new unique individual name not used before)
  - $L(x) := (L(x) \setminus \{(a : \exists R.C, I)\}) \cup X$
- Combines  $\exists$ - and  $\forall$ -rules for classical tableau systems
- Observe the updating of the index sets

# Expansion Rules

## 6. $\perp$ -rule

- If  $(a : A, I) \in L(x)$  and  $(a : \neg A, J) \in L(x)$  then
  - For every  $i \in I \cup J$  do:
    - Create a new child  $y$  of  $x$
    - $L(y) := L(x) \setminus \{(b : D, K) \mid i \in K\}$ ;  $E(y) := E(x) \cup \{i\}$
- This rule breaks clashes one by one
- Observe the updating of the exclusion-sets

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# Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Initialise root node

$$L(r) = \{(a : A_1, \emptyset)\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $D^+$ -rule to  $A_1$

$$L(r) = \{(a : A_1, \emptyset)\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $D^+$ -rule to  $A_1$

$$L(r) = \{(a : A_1, \emptyset), (a : (A_2 \sqcap A_3 \sqcap A_4, \{1\}))\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $\sqcap$ -rule twice:

$$L(r) = \{(a : A_1, \emptyset), (a : (A_2 \sqcap A_3 \sqcap A_4, \{1\}))\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $\sqcap$ -rule twice:

$$L(r) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\})\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $D^+$ -rule to  $A_2$

$$L(r) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\})\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $D^+$ -rule to  $A_2$

$$L(r) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\}), \\ (a : \forall R.D, \{1, 2\})\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $D^+$ -rule to  $A_3$

$$L(r) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\}), \\ (a : \forall R.D, \{1, 2\})\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $D^+$ -rule to  $A_3$

$$L(r) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\}), \\ (a : \forall R.D, \{1, 2\}), (a : \exists R.C, \{1, 3\})\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $D^+$ -rule to  $A_4$

$$L(r) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\}), \\ (a : \forall R.D, \{1, 2\}), (a : \exists R.C, \{1, 3\})\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $D^+$ -rule to  $A_4$

$$L(r) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\}), \\ (a : \forall R.D, \{1, 2\}), (a : \exists R.C, \{1, 3\}), (a : \forall R.\neg D, \{1, 4\})\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $\exists$ -rule

$$L(r) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\}), \\ (a : \forall R.D, \{1, 2\}), (a : \exists R.C, \{1, 3\}), (a : \forall R.\neg D, \{1, 4\})\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

Apply the  $\exists$ -rule

$$L(r) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\}), \\ (a : \forall R.D, \{1, 2\}), (a : \exists R.C, \{1, 3\}), (a : \forall R.\neg D, \{1, 4\}), \\ (b : D, \{1, 2, 3\}), (b : C, \{1, 3\}), (b : \neg D, \{1, 3, 4\})\}$$

## Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

$\perp$ -rule: New child for each element of  $\{1, 2, 3\} \cup \{1, 3, 4\}$

$L(r) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\}),$

$(a : \forall R.D, \{1, 2\}), (a : \exists R.C, \{1, 3\}), (a : \forall R.\neg D, \{1, 4\}),$

$(b : D, \{1, 2, 3\}), (b : C, \{1, 3\}), (b : \neg D, \{1, 3, 4\})\}$

# Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

$\perp$ -rule: New child for 1

$$L(x_1) = \{(\cancel{a : A_1, \emptyset}), (\cancel{a : A_2, \{1\}}), (\cancel{a : A_3, \{1\}}), (\cancel{a : A_4, \{1\}}),$$

$$(\cancel{a : \forall R.D, \{1, 2\}}), (\cancel{a : \exists R.C, \{1, 3\}}), (\cancel{a : \forall R.\neg D, \{1, 4\}}),$$

$$(\cancel{b : D, \{1, 2, 3\}}), (\cancel{b : C, \{1, 3\}}), (\cancel{b : \neg D, \{1, 3, 4\}})\}$$

# Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

$\perp$ -rule: New child for 2

$$L(x_2) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\}), \\ (a : \forall R.D, \{1, 2\}), (a : \exists R.C, \{1, 3\}), (a : \forall R.\neg D, \{1, 4\}), \\ (b : D, \{1, 2, 3\}), (b : C, \{1, 3\}), (b : \neg D, \{1, 3, 4\})\}$$

# Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

$\perp$ -rule: New child for 3

$$L(x_3) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\}), \\ (a : \forall R.D, \{1, 2\}), (a : \exists R.C, \{1, 3\}), (a : \forall R.\neg D, \{1, 4\}), \\ (b : D, \{1, 2, 3\}), (b : C, \{1, 3\}), (b : \neg D, \{1, 3, 4\})\}$$

# Example

$$\Gamma = \{A_1 \doteq A_2 \sqcap A_3 \sqcap A_4, A_2 \doteq \forall R.D, A_3 \doteq \exists R.C, A_4 \doteq \forall R.\neg D\}$$

$\perp$ -rule: New child for 4

$$L(x_4) = \{(a : A_1, \emptyset), (a : A_2, \{1\}), (a : A_3, \{1\}), (a : A_4, \{1\}), \\ (a : \forall R.D, \{1, 2\}), (a : \exists R.C, \{1, 3\}), (a : \forall R.\neg D, \{1, 4\}), \\ (b : D, \{1, 2, 3\}), (b : C, \{1, 3\}), (b : \neg D, \{1, 3, 4\})\}$$

# Algorithm Properties

- Sound, Complete, Terminating
- Checking  $A_j$ -concept satisfiability is PSPACE-complete
- Maintenance of index-sets yields EXPTIME
- Preliminary experiments give faster runtimes than a black box approach using FACT++
- Easily modified to find MSSs for all concept names

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# Future Work

## Belief Change

- How to reconcile new information with existing beliefs
  - Revision: Add a new statement to your current beliefs
  - Contraction: Remove a statement from your current beliefs

## Ontology Change

- During ontology construction the subsumption hierarchy can be computed automatically
- In practice ontology engineers want to modify this
- Add a subsumption link explicitly: revision
- Remove a subsumption link explicitly: contraction