

Computing Inconsistency Measurements under Multi-Valued Semantics by Partial Max-SAT Solvers

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Outline

- ▶ Motivation
- ▶ Inconsistency Measures under Multi-Valued Semantics
- ▶ Relationship among Different Measurements
- ▶ Encoding Algorithms
- ▶ Conclusion and Future Work

Motivation

- ▶ Consistent KBs serve as useful knowledge resources v.s. inconsistent KBs imply any conclusion (meaningless!)
- ▶ For handling inconsistent KBs:
 - paraconsistent reasoning (1960s)
 - knowledge diagnose and repair (1980s)
 - Which approach should we take?
- ↪ inconsistency measurement: a guidance to choose different approaches (2000s)

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 - Which approach should we take?
- ↪ inconsistency measurement: a guidance to choose different approaches (2000s)
- ▶ Problem
 - Relationship among different measurements
 - Efficient algorithms

Definitions

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► Multi-Valued Semantics

- 4-valued, 3-valued, LP_m , Quasi-Classical, ...
- $I : \text{Var}(K) \rightarrow \{t, f, \text{Both}, \text{None}\}$

Definitions

- ▶ Multi-Valued Semantics
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 - $I : \text{Var}(K) \rightarrow \{t, f, \text{Both}, \text{None}\}$
- ▶ ID of K respect to I under i -semantics ($i = 3, 4, LP_m, Q$)

$$ID_i(K, I) = \frac{|\{p \mid p^I = \text{Both}, p \in \text{Var}(K)\}|}{|\text{Var}(K)|}, \text{ if } I \models_i K$$

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$$ID_i(K, I) = \frac{|\{p \mid p^I = \text{Both}, p \in Var(K)\}|}{|Var(K)|}, \text{ if } I \models_i K$$

- ▶ ID of K under under i -semantics ($i = 3, 4, LP_m, Q$)

$$ID_i(K) = \min_{I \models K} ID_i(K, I)$$

Inconsistency Degree under 4-valued Semantics

Truth values: $\{t, f, B, N\}$

4-model I :

$K \rightarrow \{t, B\}$

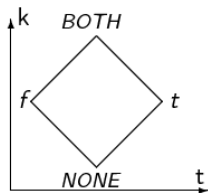


Figure: FOUR

$$\blacktriangleright ID_4(K, I) = \frac{|\{p \mid p^I = B, p \in \text{Var}(K)\}|}{|\text{Var}(K)|}$$

$$ID_4(K) = \min_{I \models_4 K} ID_4(K, I),$$

$$\rightsquigarrow K = \{p, \neg q, \neg p \vee q, r \vee s\}$$

$$\rightsquigarrow I_1 : p^{I_1} = B, q^{I_1} = f, r^{I_1} = t, s^{I_1} = t,$$

$$I_2 : p^{I_2} = B, q^{I_2} = B, r^{I_2} = t, s^{I_2} = t$$

$$I_3 : p^{I_3} = B, q^{I_3} = B, r^{I_3} = t, s^{I_3} = N$$

$$\rightsquigarrow ID_4(K, I_1) = \frac{1}{4}, ID_4(K, I_2) = \frac{2}{4}$$

$$ID_4(K, I_3) = \frac{2}{4}$$

$$ID_4(K) = \frac{1}{4}$$

Inconsistency Degree under 3-valued Semantics

Truth values: $\{t, f, B\}$

3-model I :

$K \rightarrow \{t, B\}$

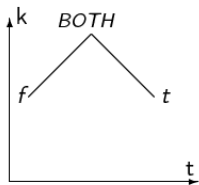


Figure: Three

$$\blacktriangleright ID_3(K, I) = \frac{|\{p | p^I = B, p \in \text{Var}(K)\}|}{|\text{Var}(K)|}$$

$$ID_3(K) = \min_{I \models_3 K} ID_3(K, I),$$

$$\rightsquigarrow K = \{p, \neg q, \neg p \vee q, r \vee s\}$$

$$\rightsquigarrow I_1 : p^{I_1} = B, q^{I_1} = f, r^{I_1} = t, s^{I_1} = t$$

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Inconsistency Degree under LPm Semantics

$$\begin{aligned} \blacktriangleright ID_{LP_m}(K, I) &= \frac{|\{p \mid p^I = B, p \in \text{Var}(K)\}|}{|\text{Var}(K)|} \\ ID_{LP_m}(K) &= \min_{I \models_{LP_m} K} ID_{LP_m}(K, I), \end{aligned}$$

LP_m interpretation:

- ▶ 3-valued interpretation
- ▶ only “most classical” ones are considered

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$$ID_{LP_m}(K) = \frac{1}{4}$$

Inconsistency Degree under Quasi-Classical Semantics

Quasi-Classical (Q)
interpretation:

► 4-valued interpretation

► Resolution laws are
satisfied

$$I \models_Q \alpha \vee \beta,$$

$$I \models_Q \neg\beta \vee \gamma$$

$$\Rightarrow I \models_Q \alpha \vee \gamma$$

$$\blacktriangleright ID_Q(K, I) = \frac{|\{p \mid p^I = B, p \in \text{Var}(K)\}|}{|\text{Var}(K)|}$$

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Theorem

Given a knowledge base K , then

$$ID_3(K) = ID_4(K) = ID_{LP_m}(K) \leq ID_Q(K).$$

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- ▶ $ID_3(K) \geq ID_4(K)$: Trivial since $I \models_3 K \Rightarrow I \models_4 K$

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- ▶ $ID_3(K) \leq ID_4(K)$: Every 4-model can be modified to a 3-model by changing N to t while preserving the inconsistency degree.

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- ▶ $ID_3(K) \geq ID_{LP_m}(K)$: Assume that $ID_3(K) < ID_{LP_m}(K)$. Then we can find a contradiction.

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- ▶ $ID_3(K) \leq ID_{LP_m}(K)$: Trivial since $I \models_{LP_m} K \Rightarrow I \models_3 (K)$.

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- ▶ $ID_3(K) \geq ID_{LP_m}(K)$: Assume that $ID_3(K) < ID_{LP_m}(K)$. Then we can find a contradiction.
- ▶ $ID_3(K) \leq ID_{LP_m}(K)$: Trivial since $I \models_{LP_m} K \Rightarrow I \models_3 (K)$.
- ▶ $ID_4(K) \leq ID_Q(K)$: Trivial since $I \models_Q K \Rightarrow I \models_4 (K)$.

Partial Max-SAT Problem

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► Partial Max-SAT Problem:

- Optimized Version of SAT problem
- $P = (H, S)$
- H : hard clauses, all must be satisfied
- S : soft clauses, should be satisfied as many as possible
- $\hat{I} = \arg \max_I |\{\gamma \mid \gamma \in S, I \models \gamma, I \models H\}|$.

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▶ Max-SAT Competition

- <http://www.maxsat.udl.cat/09/>
- <http://www.maxsat.udl.cat/10/>

Computing Inconsistency Degrees

- ▶ only KB as a set of clauses (CNF) considered
- ▶ consider ID_4 and ID_Q (Since $ID_3(K) = ID_4(K) = ID_{LP_m}(K) \leq ID_Q(K)$)

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Road Map

- 1 Multi-valued semantics \Rightarrow 2-valued semantics.
- 2 Represent ID_i by 2-valued semantics.
- 3 $ID_i \Rightarrow$ partial Max-SAT problem.

4-valued Logic to 2-valued Logic

$$\begin{array}{ll} K = \{\gamma_1, \gamma_2, \dots, \gamma_n\} & 4(K) = \{4(\gamma_1), 4(\gamma_2), \dots, 4(\gamma_n)\} \\ \gamma = l_1 \vee \dots \vee l_k & \Rightarrow 4(\gamma) = 4(l_1) \vee \dots \vee 4(l_k) \\ l = p & 4(p) = +p \\ l = \neg p & 4(\neg p) = -p \end{array}$$

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Example

$$K = \{\neg p, p \vee q, \neg q, r\} \quad \Rightarrow \quad 4(K) = \{-p, +p \vee +q, -q, +r\}$$

4-valued Logic to 2-valued Logic

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Example

$$K = \{\neg p, p \vee q, \neg q, r\} \quad \Rightarrow \quad 4(K) = \{-p, +p \vee +q, -q, +r\}$$

Remark

- ▶ $4(K)$ is a knowledge base over variables $\{+p, -p \mid p \in \text{Var}(K)\}$
- ▶ A 4-valued interpretation I can also be seen as a 2-valued interpretation on $\{+p, -p \mid p \in \text{Var}(K)\}$.

4-valued Logic to 2-valued Logic

Theorem ([?])

$$I \models_4 K \Leftrightarrow I \models 4(K)$$

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Example

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4-valued Logic to 2-valued Logic

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Example

- ▶ $K = \{\neg p, p \vee q, \neg q, r\} \Rightarrow 4(K) = \{-p, +p \vee +q, -q, +r\}$
- ▶ $I_1 = \{+p, -p, -q, +r\}$

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Example

- ▶ $K = \{\neg p, p \vee q, \neg q, r\} \Rightarrow 4(K) = \{-p, +p \vee +q, -q, +r\}$
- ▶ $I_1 = \{+p, -p, -q, +r\}$
- ▶ As 4-interpretation over $\{p, q, r\}$:
 $p^{I_1} = B, q^{I_1} = f, r^{I_1} = t.$

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Theorem ([?])

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- ▶ As 4-interpretation over $\{p, q, r\}$:
 $p^{I_1} = B, q^{I_1} = f, r^{I_1} = t.$
- ▶ As 2-interpretation over $\{+p, -p, +q, -q, +r, -r\}$:
 $+p^{I_1} = t, -p^{I_1} = t, -q^{I_1} = t, +r^{I_1} = t, +q^{I_1} = f, -r^{I_1} = f.$

4-valued Logic to 2-valued Logic

Theorem ([?])

$$I \models_4 K \Leftrightarrow I \models 4(K)$$

Example

- ▶ $K = \{\neg p, p \vee q, \neg q, r\} \Rightarrow 4(K) = \{-p, +p \vee +q, -q, +r\}$
- ▶ $I_1 = \{+p, -p, -q, +r\}$
- ▶ As 4-interpretation over $\{p, q, r\}$:
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- ▶ As 2-interpretation over $\{+p, -p, +q, -q, +r, -r\}$:
 $+p^{I_1} = t, -p^{I_1} = t, -q^{I_1} = t, +r^{I_1} = t, +q^{I_1} = f, -r^{I_1} = f.$
- ▶ We have $I_1 \models_4 K$ and $I_1 \models 4(K).$

Representing ID_4 by 2-valued logic

$$ID_4(K, I) = \frac{|\{p \mid p^I = B, p \in \text{Var}(K)\}|}{|\text{Var}(K)|},$$

$$ID_4(K) = \min_{I \models_4 K} ID_4(K, I)$$

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↓

$$ID_4(K, I) = \frac{|\{p \mid +p^I = t \wedge -p^I = t, p \in \text{Var}(K)\}|}{|\text{Var}(K)|};$$

$$ID_4(K) = \min_{I \models_4(K)} ID_4(K, I).$$

$ID_4 \Rightarrow$ Partial Max-SAT Problem

$$\begin{aligned} & \min_{I \models 4(K)} |\{p \mid p \in \text{Var}(K), +p^I = t \wedge -p^I = t\}| \\ & = \min_{I \models 4(K)} |\{p \mid p \in \text{Var}(K), (\neg +p \vee \neg -p)^I = f\}| \\ & = \max_{I \models 4(K)} |\{p \mid p \in \text{Var}(K), (\neg +p \vee \neg -p)^I = t\}|. \end{aligned}$$

$I \models 4(K) \Rightarrow$ Hard constraints

$\max|\dots| \Rightarrow$ Soft Constraints

$ID_4 \Rightarrow$ Partial Max-SAT Problem

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$I \models 4(K) \Rightarrow$ Hard constraints

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Definition

Given $K = \{\gamma_1, \dots, \gamma_n\}$, the corresponding partial Max-SAT problem for ID_4 , written $P_4(K) = (H_4(K), S_4(K))$, is defined as follows:

$$H_4(K) = \{4(\gamma) \mid \gamma \in K\};$$

$$S_4(K) = \{\neg +p \vee \neg -p \mid p \in \text{Var}(K)\}.$$

$ID_4 \Rightarrow$ Partial Max-SAT Problem

Theorem

Suppose I is a solution to the partial Max-SAT problem $P_4(K)$. Let $b = |\{p \mid +p^I = t \wedge -p^I = t\}|$ and $m = |\text{Var}(K)|$. Then $ID_4(K) = b/m$.

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Example

► $K = \{\neg p, p \vee q, \neg q, r\}$

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- ▶ $P_4(K) = (H_4(K), S_4(K))$
 $H_4(K) = \{-p, +p \vee +q, -q, +r\}$
 $S_4(K) = \{\neg + p \vee \neg - p, \neg + q \vee \neg - q, \neg + r \vee \neg - r\}$

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- ▶ One optimized solution I :
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- ▶ Corresponding 4-model : $p^I = B, q^I = f, r^I = t$.

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- ▶ $P_4(K) = (H_4(K), S_4(K))$
 $H_4(K) = \{-p, +p \vee +q, -q, +r\}$
 $S_4(K) = \{\neg +p \vee \neg -p, \neg +q \vee \neg -q, \neg +r \vee \neg -r\}$
- ▶ One optimized solution I :
 $+p^I = t, -p^I = t, +q^I = f, -q^I = t, +r^I = t, -r^I = f$.
- ▶ Corresponding 4-model : $p^I = B, q^I = f, r^I = t$.
- ▶ $ID_4(K) = 1/3$

Algorithm 1 Computing ID_4 by Partial MAX-SAT Solver

```
1: procedure  $ID_4(K)$ 
2:    $P \leftarrow \{\}$ 
3:    $m \leftarrow |Var(K)|$ 
4:   for all Clause  $\gamma \in K$  do
5:      $P.addHardClause(4(\gamma))$ 
6:   end for
7:   for all Variable  $p \in Var(K)$  do
8:      $P.addSoftClause(\neg + p \vee \neg - p)$ 
9:   end for
10:   $l \leftarrow PartialMaxSATSolver(P)$ 
11:   $b = |\{p | +p^l = t \wedge -p^l = t\}|$ 
12:  return  $b/m$ 
13: end procedure
```

Computing ID_Q

- 1 QC semantics \Rightarrow 2-valued semantics.
- 2 Represent ID_Q by 2-valued semantics.
- 3 $ID_Q \Rightarrow$ partial Max-SAT problem.

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Problem: $Q(l_1 \vee \dots \vee l_n)$ can not keep CNF!

Solution: Introduce new variables and convert it to CNF

Algorithm 2 Computing ID_Q by Partial Max-SAT Solver

```
1: procedure  $ID_Q(K)$ 
2:    $P \leftarrow \{\}$ 
3:    $m \leftarrow |Var(K)|$ 
4:   for all Clause  $\gamma = \{l_1, \dots, l_n\} \in K$  do
5:      $P.addHardClause(y_1 \vee \dots \vee y_n \vee z)$ 
6:     for  $i = 1$  to  $n$  do
7:        $P.addHardClause(\neg y_i \vee +l_i)$ 
8:        $P.addHardClause(\neg y_i \vee \neg -l_i)$ 
9:        $P.addHardClause(\neg z \vee +l_i)$ 
10:       $P.addHardClause(\neg z \vee -l_i)$ 
11:    end for
12:  end for
13:  for all  $p \in Var(K)$  do
14:     $P.addSoftClause(\neg +p \vee \neg -p)$ 
15:  end for
16:   $I \leftarrow \text{PartialMaxSATSolver}(P)$ 
17:   $b = |\{p \mid +p^I = t \wedge -p^I = t\}|$ 
18:  return  $b/m$ 
19: end procedure
```


Evaluation

► Data Set:

- SAT benchmark SATLIB <http://www.satlib.org>
- automotive product configuration

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► Partial Max-SAT Solvers:

- SAT4j MaxSAT
- MsUncore
- Clone

Instance				Encoding Algorithm		
name	#V	#C	ID_4	sat4j	MsUncore	clone
uuf50-0101	50	218	0.02000	0.396	0.026	1.119
uuf50-0102	50	218	0.02000	0.398	0.020	1.121
uuf50-0103	50	218	0.02000	0.450	0.044	1.142
uuf50-0104	50	218	0.02000	0.397	0.027	1.279
uuf75-011	75	325	0.01330	0.496	0.031	1.379
uuf75-012	75	325	0.01330	0.447	0.030	1.355
uuf75-013	75	325	0.01330	0.443	0.033	1.333
uuf75-014	75	325	0.01333	0.494	0.029	1.372
uuf100-0101	100	430	0.01000	0.545	0.045	1.748
uuf100-0102	100	430	0.01000	0.918	0.053	2.088
uuf100-0103	100	430	0.02000	3.951	2.592	*
C168_FW_SZ_107	1698	5401	0.00059	0.698	0.120	*
C168_FW_SZ_128	1698	5422	0.00059	0.601	0.090	13.191
C168_FW_SZ_41	1698	7489	0.00059	0.849	0.085	11.939

Table: Computing ID_4 by Encoding Algorithm

Instance				Encoding Algorithm		
name	#V	#C	ID_Q	sat4j	MSUnCore	clone
uuf50-0101	50	218	1.000	0.445	*	0.428
uuf50-0102	50	218	1.000	0.444	*	0.446
uuf50-0103	50	218	1.000	0.449	*	0.246
uuf50-0104	50	218	1.000	0.494	*	0.433
uuf75-011	75	325	1.000	0.544	*	0.434
uuf75-012	75	325	1.000	0.548	*	0.435
uuf75-013	75	325	1.000	0.455	*	1.338
uuf75-014	75	325	1.000	0.646	*	0.437
uuf100-0101	100	430	1.000	0.709	*	0.478
uuf100-0102	100	430	1.000	0.803	*	0.438
uuf100-0103	100	430	1.000	0.749	*	0.445
C168_FW_SZ_107	1698	5401	0.124	9.269	*	1.487
C168_FW_SZ_128	1698	5422	0.107	9.916	*	0.792
C168_FW_SZ_41	1698	7489	0.117	13.627	*	0.738

Table: Computing ID_Q by Encoding Algorithm

Conclusion & Future Work

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Conclusion:

- $ID_4(K) = ID_{LP_m}(K) = ID_3(K) \leq ID_Q(K)$
- $ID_4 \Rightarrow$ Partial Max-SAT
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Conclusion & Future Work

Conclusion:

- $ID_4(K) = ID_{LP_m}(K) = ID_3(K) \leq ID_Q(K)$
- $ID_4 \Rightarrow$ Partial Max-SAT
- $ID_Q \Rightarrow$ Partial Max-SAT

Future Work:

- approximating inconsistency degrees
- Other encoding: Pseudo Problem for ID_Q
- Measure inconsistent Description Logic and Logic Program.

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Questions?