

# Computing Smallest MUSes of Quantified Boolean Formulas

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# Motivation: Minimal Unsatisfiable Subsets (MUSes)

- Assuming monotonicity: **minimal explanations** as minimal sets of formulas  $S$  implying a consequence  $p$ 
  - Relation to inconsistency:  $S \rightarrow p$  is satisfiable iff  $S \wedge \neg p$  is unsatisfiable
- Propositional logic: variety of **algorithms for computing minimal unsatisfiable subsets (MUSes)** (Marques-Silva & Mencía, 2020)
  - algorithms for smallest MUSes (Liffiton et al., 2009; Ignatiev et al., 2016, 2015)
    - corresponding decision problem  $\Sigma_2^P$ -complete (Liberatore, 2005)
  - MUSes of quantified Boolean formulas (QBFs) (Lonsing & Egly, 2015)

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# Motivation: Strong Explanations

- Nonmonotonic logics: if  $S$  implies  $p$  then  $S' \supset S$  might not imply  $p$ 
  - e.g. abstract argumentation: if  $a$  is credulously accepted in  $F = (A, R)$ , then  $a$  is rejected in  $(A \cup \{d\}, R \cup \{d \rightarrow a\})$  for  $d \notin A$
- **Strong explanations**: generalization to **nonmonotonic reasoning**
  - based on **strong inconsistency**: require that  $S \subseteq K$  remains inconsistent for each  $S'$  with  $S \subseteq S' \subseteq K$  (Brewka et al., 2019)
  - Reiter's **hitting set duality** satisfied: (minimal) explanations as hitting sets of (minimal) diagnoses and vice versa
  - instantiations: answer set programming, **abstract argumentation** (Brewka & Ulbricht, 2019; Mencía & Marques-Silva, 2020; Ulbricht & Wallner, 2021)

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# Motivation: Abstract Argumentation

- Strong explanations as induced subgraphs: acceptance status of argument unchanged no matter which arguments from original argumentation framework are added (Ulbricht & Wallner, 2021)
  - subset-minimality or smallest cardinality desirable
- **Declarative approaches** to extracting smallest strong explanations for credulous rejection:  $\Sigma_2^P$ -complete problem
  - answer set programming (Saribatur et al., 2020)
  - propositional SMUS extractors (Niskanen & Järvisalo, 2020)
- **Strong explanations for credulous acceptance** under admissible and stable semantics
  - verification of a strong explanation is  $\Pi_2^P$ -complete (Ulbricht & Wallner, 2021)
  - computing smallest explanations is (clearly) in  $\Sigma_3^P$

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# Contributions

- 1 **Encodings** for computing strong explanations in abstract argumentation
  - smallest strong explanations as smallest MUSes of QBFs
  - for credulous acceptance and skeptical rejection
- 2 **Complexity** of computing smallest MUSes of QBFs
  - for  $k$ -QBFs:  $\Sigma_{k+1}^P$ -complete (leading  $\exists$ ) or  $\Sigma_k^P$ -complete (leading  $\forall$ )
- 3 **Algorithm** for computing smallest MUSes of QBFs
  - based on the implicit hitting set (IHS) approach
  - employs modern QBF solving techniques
- 4 **Implementation** of the algorithm
  - generic: allows for computing smallest MUSes of QBFs in prenex CNF
  - empirical evaluation: practical declarative approach for computing strong explanations in abstract argumentation

Implementation and benchmark data openly available:  
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# Quantified Boolean Formulas (QBFs)

Boolean satisfiability (SAT) with quantifiers:  $\exists, \forall$

- Instance:  $\Phi = \vec{Q}_k \cdot \varphi$ 
  - prefix  $\vec{Q}_k = Q_1 X_1 \cdots Q_k X_k$ :
    - alternating quantifiers  $Q_i \in \{\exists, \forall\}$ ,  $Q_i \neq Q_{i+1}$
    - pairwise disjoint sets of variables  $X_i$
  - matrix  $\varphi$ : formula over  $X_1 \cup \cdots \cup X_k$

Semantics defined recursively in a natural way:

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# QBF Optimization: Smallest MUSes of QBFs

$S^* \subset S$  is a **core** of  $\exists S \vec{Q}_k. \varphi$   
if  $\exists S \vec{Q}_k. \underbrace{\varphi [S^*]}_{= \varphi[s \mapsto T \mid s \in S^*]}$  is false.

MUSes = subset-minimal cores

Smallest MUSes = smallest-cardinality cores

*Note: definition covers all conjunctive forms, e.g. standard clausal MUSes.*

## Example

$$\begin{aligned} & \exists x_1 y_1 z_1 \forall x_2 y_2 z_2 \\ & ((x_1 \vee \neg x_2) \wedge (y_1 \vee \neg y_2) \wedge (z_1 \vee \neg z_2)) \\ & \quad \wedge (\neg x_2 \vee \neg y_2) \wedge (\neg y_2 \vee \neg z_2) \\ & \wedge (\neg x_1 \vee x_2) \wedge (\neg y_1 \vee x_2 \vee y_2) \wedge (\neg z_1 \vee y_2 \vee z_2) \rightarrow \neg z_2 \end{aligned}$$

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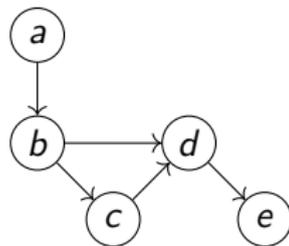
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# Strong Explanations in Abstract Argumentation

- Argumentation frameworks (AFs):  
directed graphs  $F = (A, R)$  (Dung, 1995)
- Semantics  $\sigma$  characterize jointly acceptable sets of arguments  $\sigma(F)$  called extensions
  - admissible, stable, ...
- Credulous acceptance: argument  $q \in A$  contained in an extension  $E \in \sigma(F)$ 
  - *In paper: skeptical rejection*



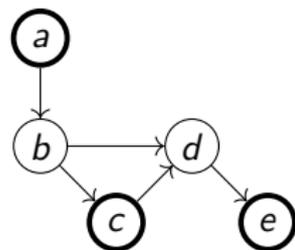
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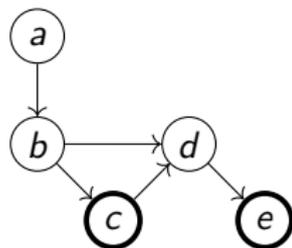
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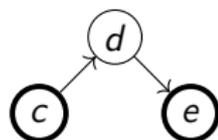
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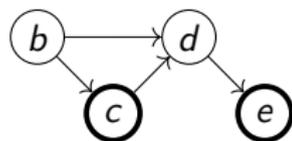
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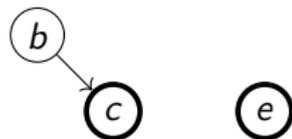
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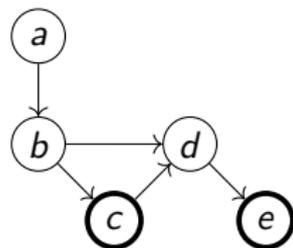
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Variables:

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→ prefix  $\exists Y \forall X$ : “ $\exists$  subframework  $\forall$  extensions”

Propositional formulas  $\varphi_\sigma(F)$  condition standard encodings of semantics  $\sigma$  on the existence of arguments: (Besnard & Doutre, 2004; Niskanen & Järvisalo, 2020)

- e.g.  $\varphi_{stb}(F) = \varphi_{cf}(F) \wedge \bigwedge_{a \in A} ((y_a \wedge \neg x_a) \rightarrow \bigvee_{(b,a) \in R} (y_b \wedge x_b))$

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# Complexity of Computing Smallest MUSes of QBFs

*How hard is it to decide whether a  $k$ -QBF admits a small core?*

Leading quantifier  $\exists$ :  $\Sigma_{k+1}^P$ -complete.

- Generalizes  $\Sigma_2^P$ -completeness for propositional logic. (Liberatore, 2005)
- Problem remains hard for DNF formulas when  $k$  is odd...
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Identify an increasing collection of *non-solutions* and *exclude them from consideration* in a minimal way.

(Moreno-Centeno & Karp, 2013; Saikko et al., 2016)

- A *correction set*  $cs \subseteq S$  renders  $\exists S \vec{Q}_k \cdot \varphi [S \setminus cs]$  true.
  - QBF solver: *extract a collection of correction sets*  $\mathcal{C}$ .
- $hs$  is a *hitting set* over  $\mathcal{C}$  if  $hs$  intersects each  $cs \in \mathcal{C}$ .
  - IP solver: *compute hitting sets with smallest cardinality*.

*Reasoning and optimization* effectively decoupled:

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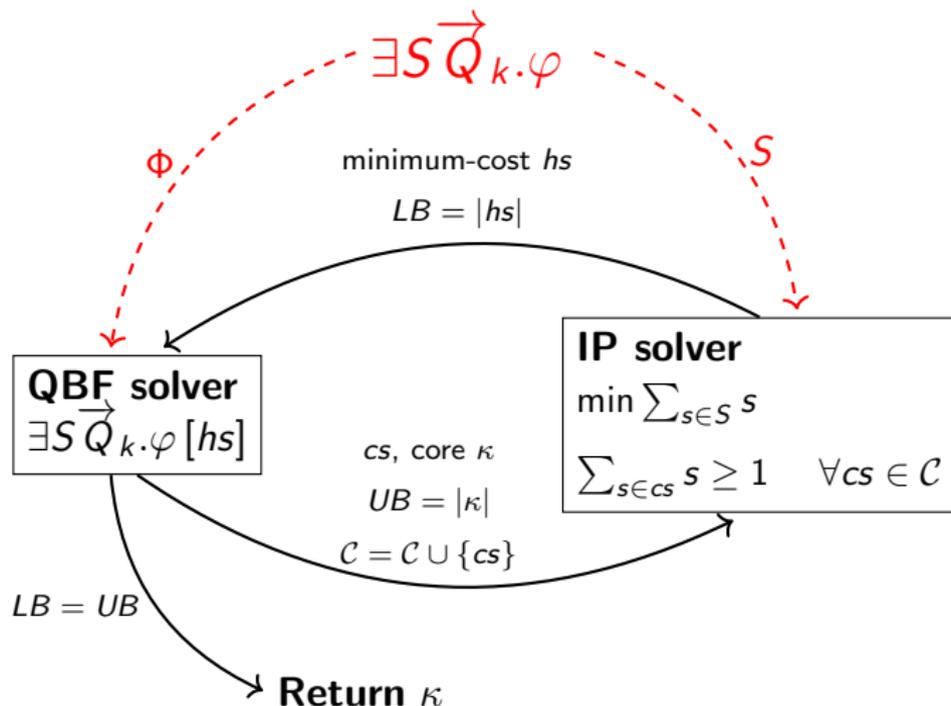
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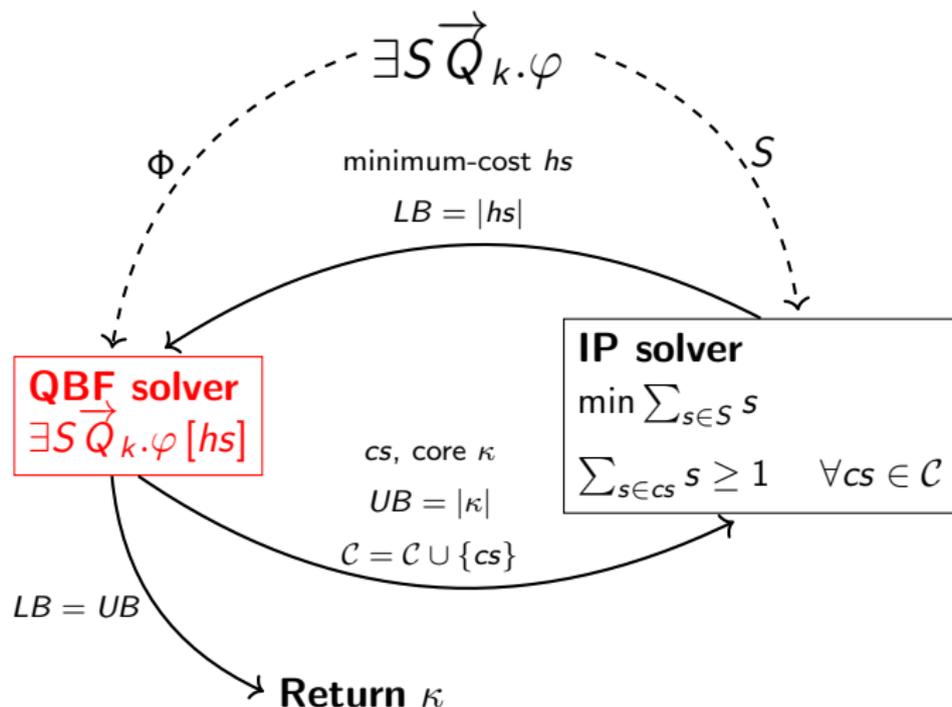
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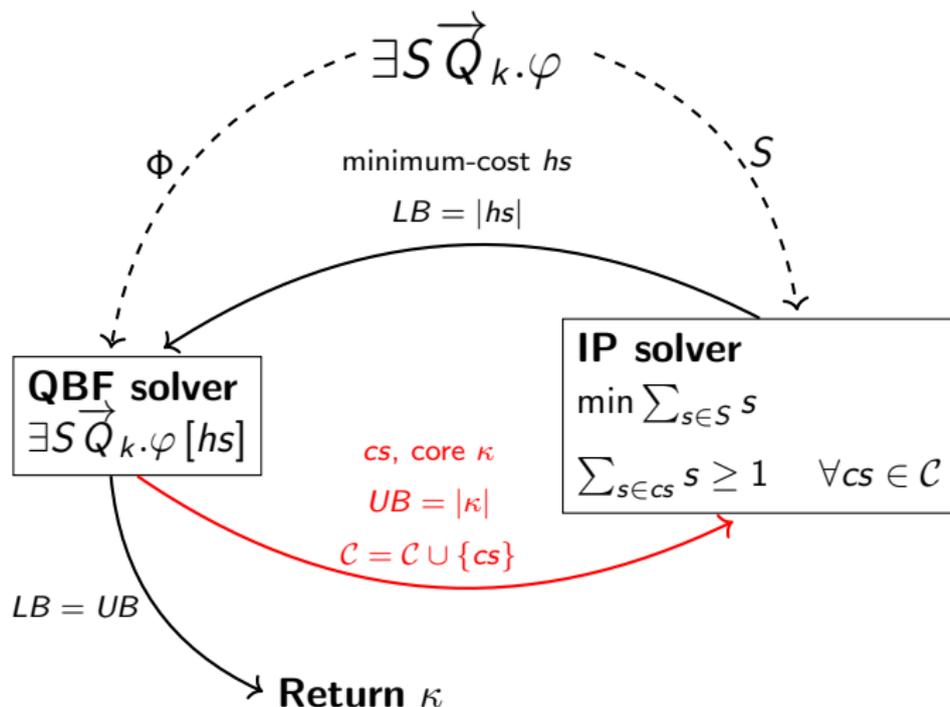
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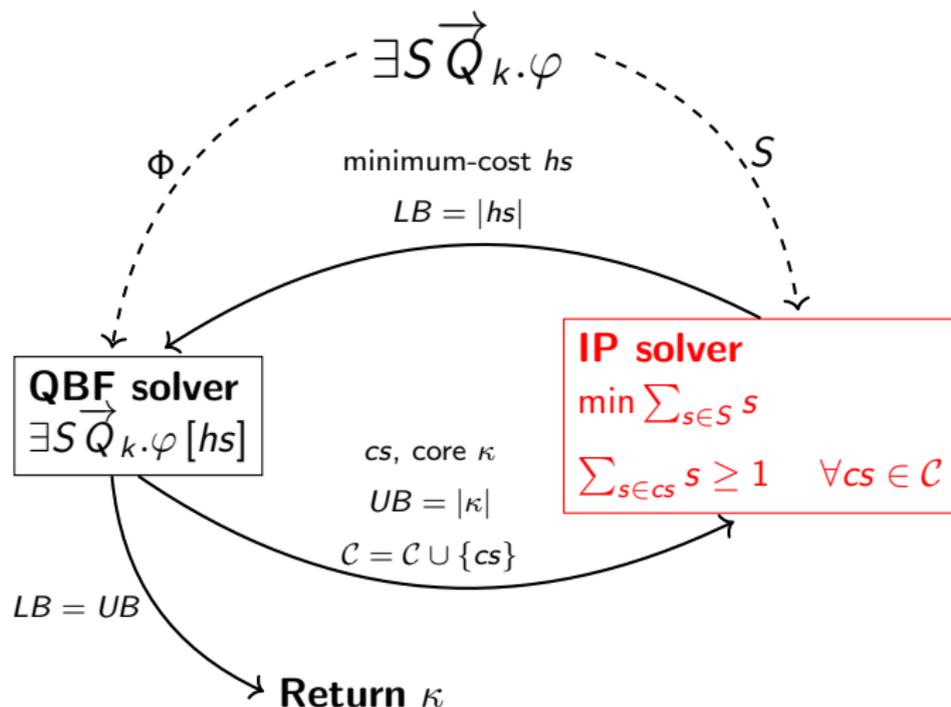
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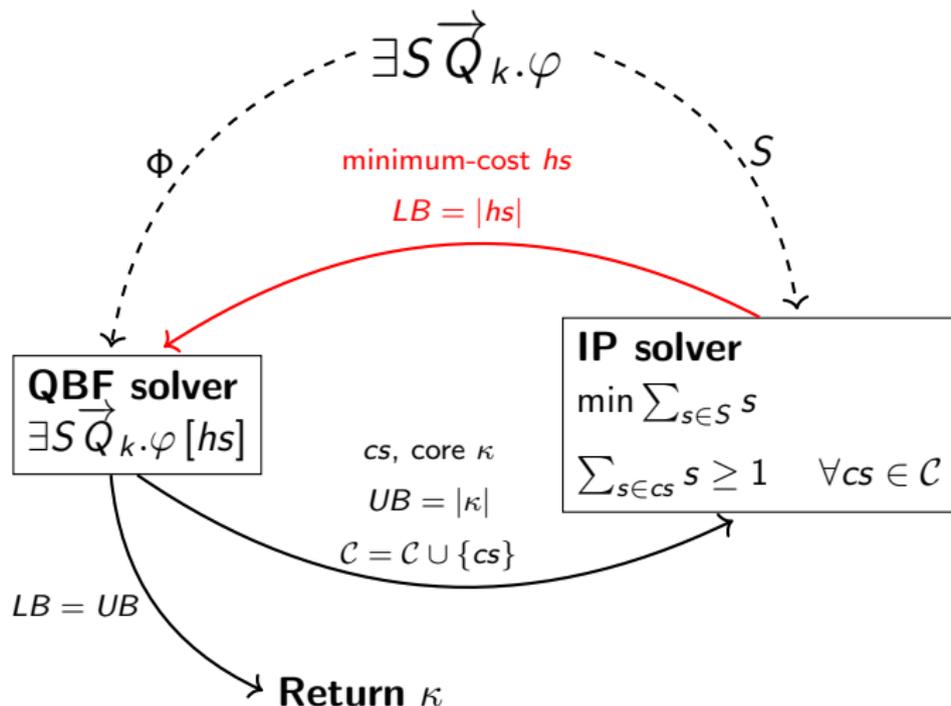
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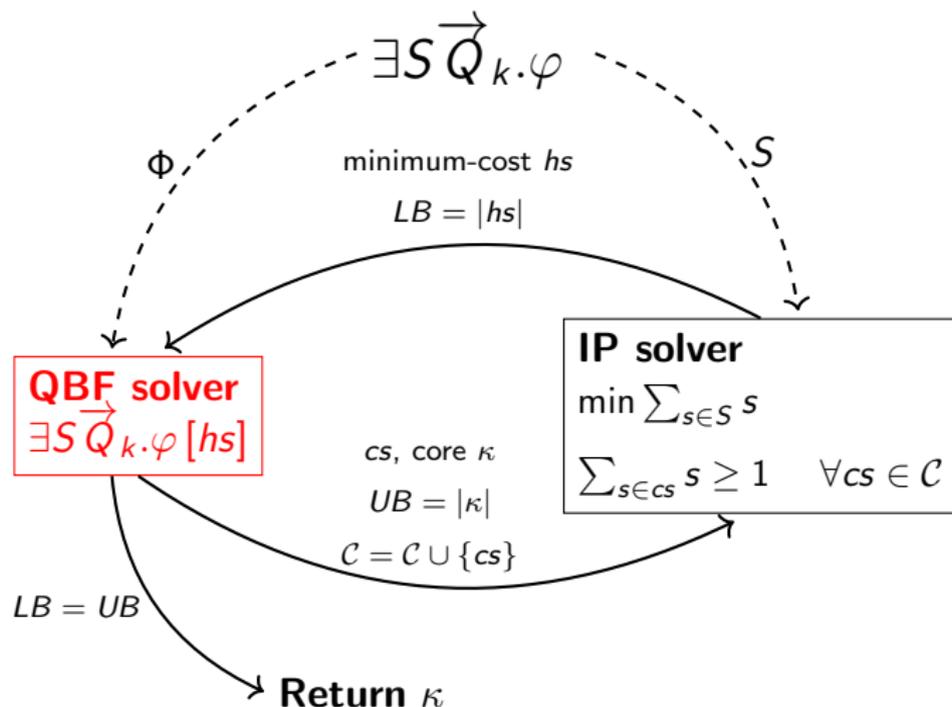
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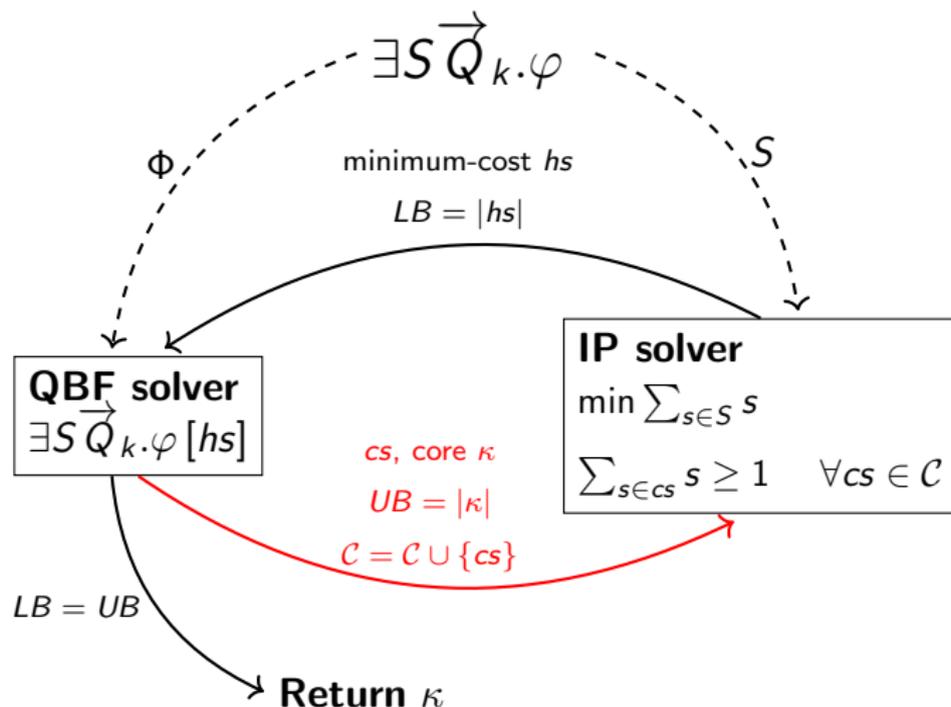
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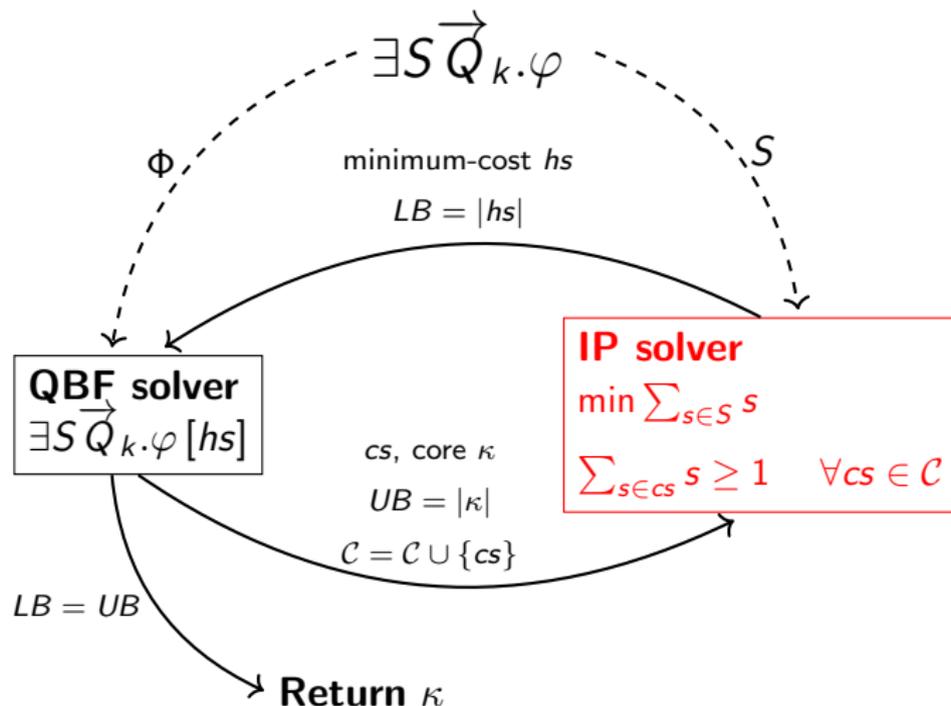
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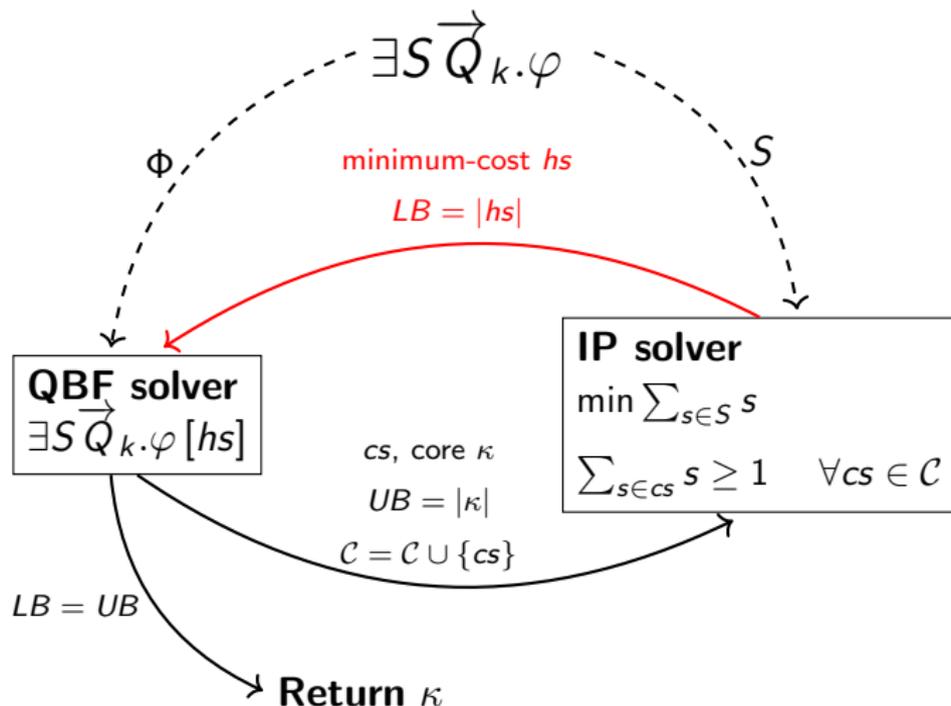
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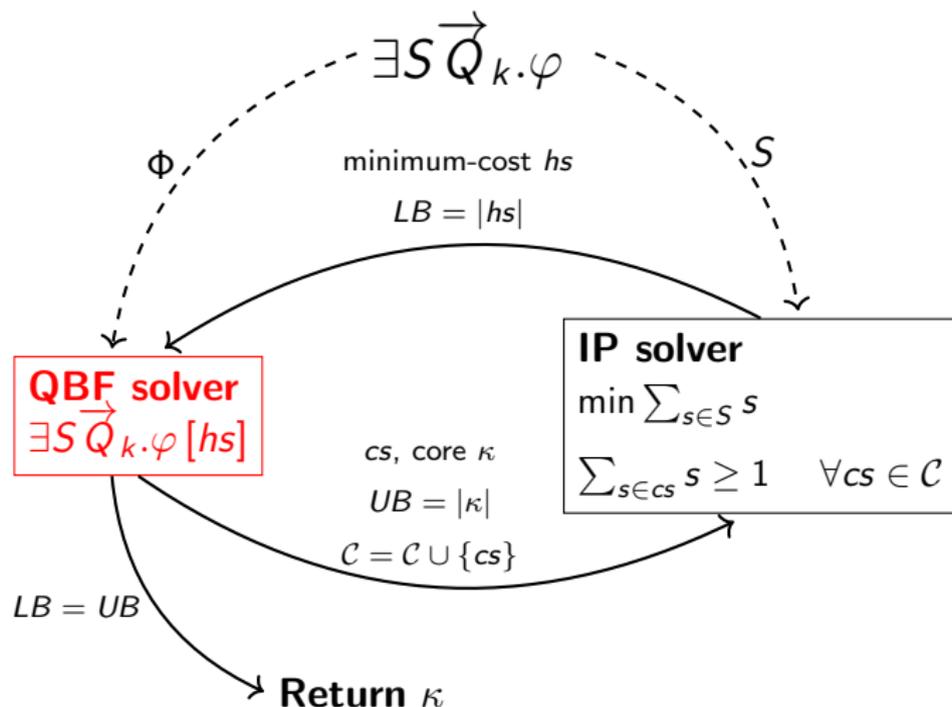
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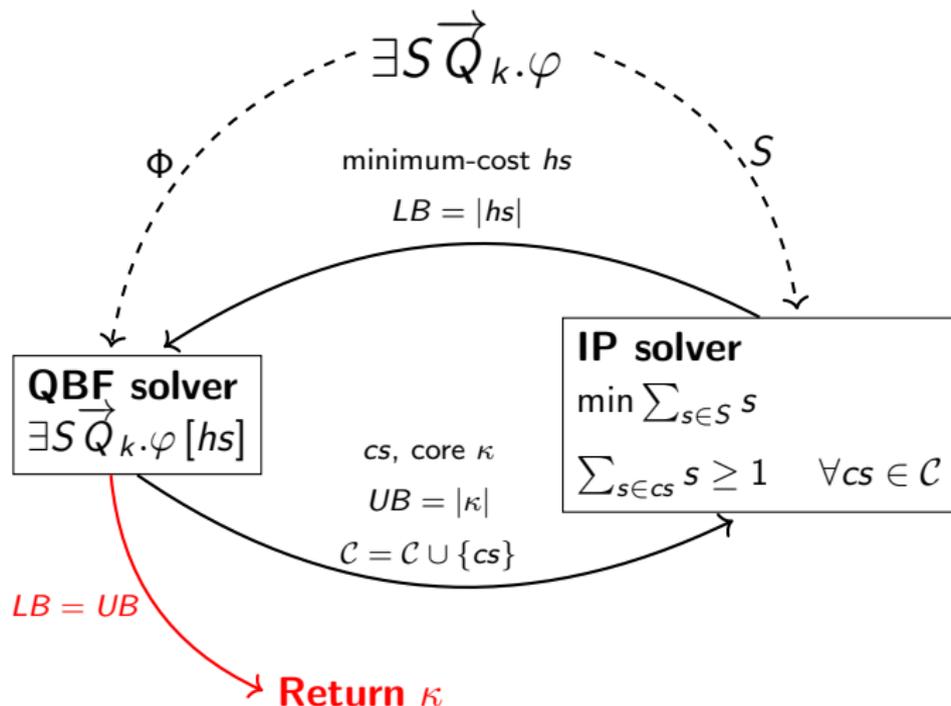
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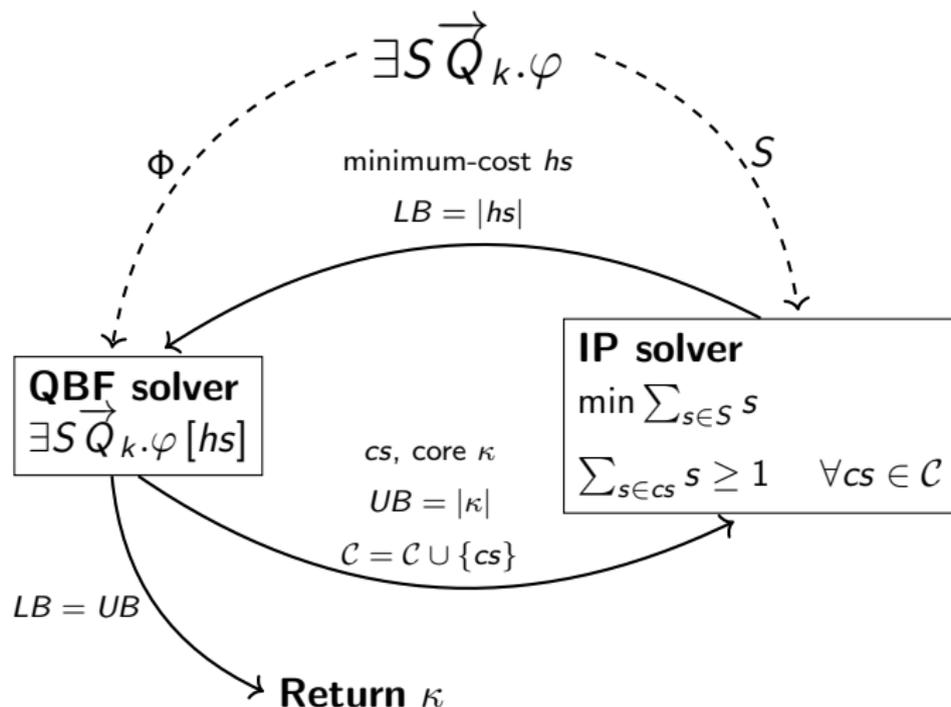
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## Benchmark Instances

- Strong explanations: all 326 AFs from ICCMA'19
  - semantics: admissible, stable
  - query arguments sampled from credulously accepted arguments
- *In paper: specific small unsatisfiable QBFLIB instances!*

## Benchmark setup

- QBF solvers: RAReQS, DepQBF
- Per-instance limits: 3600 seconds and 16 GB memory
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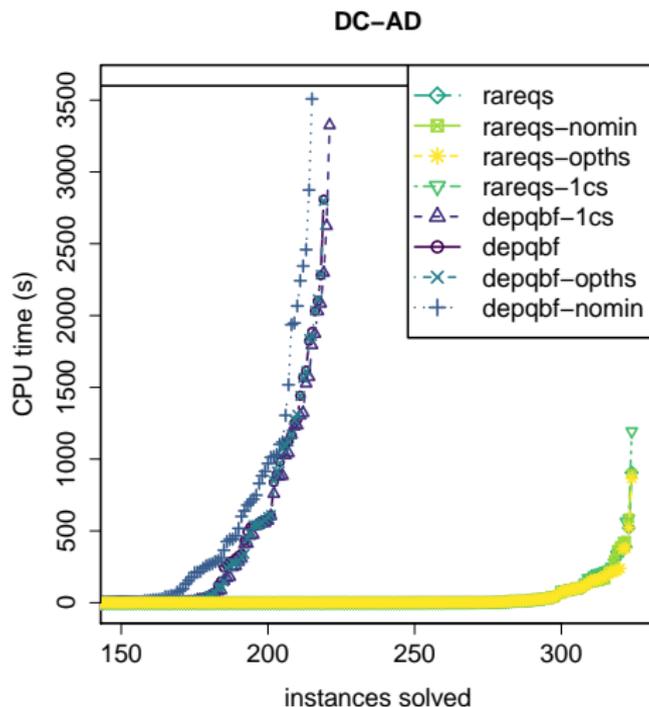
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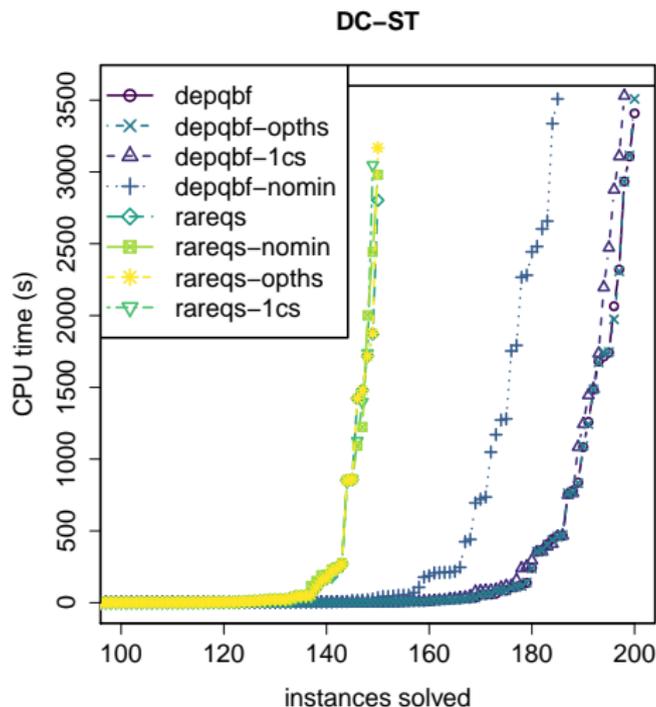
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# Empirical Evaluation



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## Contributions

A first overview on the computation of smallest MUSes of QBFs:

- Computational **complexity analysis**
  - $\Sigma_{k+1}^P$ -complete for leading existential quantifier
  - $\Sigma_k^P$ -complete for leading universal quantifier
- IHS-based **algorithm and implementation**
  - relies on iterative QBF and IP solver calls
  - additional techniques can be incorporated
- Application: **declarative encodings** for computing smallest strong explanations in abstract argumentation
  - empirical evaluation shows that the approach is viable

Implementation available online in open source:  
<https://bitbucket.org/coreo-group/qbf-smuser>

# Thank you for your attention!

Get in touch via email:  
`andreas.niskanen@helsinki.fi`

Or come chat in person :)

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## Extra: Clausal MUSes as Cores

Let  $\varphi_{CNF} = \{C_j \mid j = 1, \dots, m\} = \bigwedge_{j=1}^m C_j$  be a CNF formula.

The MUSes of  $\Phi_{PCNF} = \vec{Q}_k \cdot \varphi_{CNF}$  correspond exactly to subset-minimal cores of

$$\exists S \vec{Q}_k \cdot \bigwedge_{j=1}^m (s_j \rightarrow C_j)$$

with  $S = \{s_1, \dots, s_m\}$ : if  $S^* \subset S$  is a core, then  $\vec{Q}_k \cdot \bigwedge_{j: s_j \in S^*} C_j$  is false, and vice versa.

Same holds for smallest MUSes and smallest-cardinality cores.