

# 1 Explanation Generation

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**Algorithm 1:** CSP-Explain( $\mathcal{T}, f [, \mathcal{I}_0]$ )

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input           :  $\mathcal{T}$  set of constraints
input           :  $f$  a consistent objective function
optional input:  $\mathcal{I}_0$  a partial interpretation
output          : Explanation sequence
1 begin
2    $\mathcal{I} \leftarrow \mathcal{I}_0$  // Initial partial interpretation
3    $\mathcal{I}_{end} \leftarrow \text{propagate}(\mathcal{I}, \mathcal{T})$  // Goal state
4    $Seq \leftarrow \text{empty set}$  // explanation sequence
5   while  $\mathcal{I} \neq \mathcal{I}_{end}$  do
6      $\mathcal{F} \leftarrow \mathcal{I}_{end} \setminus \mathcal{I}$ ; // Facts to be derived
7      $\mathcal{F}' \leftarrow \{\neg \mathcal{F}\}$ ; // Set with all negated literals of  $\mathcal{F}$ 
8     //
9     // Propagate all negated literals OMUS finds the smallest explanation
10    //
11     $X \leftarrow \text{OMUS}(\mathcal{F}' \wedge \mathcal{I} \wedge \mathcal{S})$ ;
12     $E \leftarrow \mathcal{I} \cap X$ ; // Explanation used
13     $\mathcal{N} \leftarrow \text{propagate}(E \wedge \mathcal{S})$ ; // Newly derived facts
14     $\mathcal{I} \leftarrow \mathcal{I} \cup \mathcal{N}$ ; // Update known facts
15    //
16    // Add explanation for newly derived facts to explanation sequence
17    //
18    for  $n \in \mathcal{N}$  do
19      |  $(E_n, \mathcal{S}_n, n)$  to  $Seq$ ;
20    end
21  end
22 end

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## 2 OMUS Algorithm

Note that if we assign a unit weight to every element in the subset, we reduce the problem of finding an OMUS to finding a SMUS.

**Definition 1.** Let  $\Gamma$  be a collection of sets and  $HS(\Gamma)$  the set of all hitting sets on  $\Gamma$  and let  $f$  be an valid objective function. Then a hitting set  $h \in \Gamma$  is said to be an **optimal** hitting set if  $\forall h' \in HS(\Gamma)$  we have that  $f(h) \leq f(h')$ .

**Property 1.** The **optimal** hitting set of a collection of sets  $\Gamma$  is denoted by  $OHS(\Gamma)$ .

The algorithm is based on the following observation:

**Proposition 1.** A set  $\mathcal{U} \subseteq \mathcal{F}$  is an OMUS of  $\mathcal{F}$  if and only if  $\mathcal{U}$  is an optimal hitting set of  $MCSes(\mathcal{F})$

**Lemma 1.** Let  $\mathcal{K} \subseteq MCSes(\mathcal{F})$ . Then a subset  $\mathcal{U}$  of  $\mathcal{F}$  is an OMUS if  $\mathcal{U}$  is a optimal hitting set on  $\mathcal{K}$  and  $\mathcal{U}$  is unsatisfiable

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**Algorithm 2:** OMUS-Delayed( $\mathcal{F}$ ,  $[f, \mathcal{H}_0]$ )

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input      :  $\mathcal{F}$  a CNF formula
input      :  $cost$  a cost function
optional input:  $\mathcal{H}_0$  initial collection of disjoint Minimum Correction Sets
output     :  $OMUS(\mathcal{F})$ 

1 begin
2    $\mathcal{H} \leftarrow \text{DisjointMCS}(\mathcal{F})$ 
3   while true do
4      $hs \leftarrow \text{OptimalHittingSet}(\mathcal{H}, cost)$  // Find optimal solution
5      $(sat?, \mu) \leftarrow \text{SatSolver}(hs)$ 
6     // If SAT,  $\mu$  contains the satisfying truth assignment
7     // IF UNSAT,  $hs$  is the OMUS
8     if not sat? then
9       break
10    end
11     $\mathcal{C} \leftarrow \mathcal{F} \setminus \text{Grow}(hs)$ 
12     $\mathcal{H} \leftarrow \mathcal{H} \cup \{\mathcal{C}\}$ 
13     $nonOptLevel \leftarrow 0$ 
14    // Find a series of non-optimal solutions
15    while true do
16      switch nonOptLevel do
17        case 0
18          // Add/Remove clause (choose clause appears most frequently in the
19          // set of hitting sets so far)
20           $hs \leftarrow \text{FindIncrementalHittingSet}(\mathcal{H}, \mathcal{C}, hs)$ 
21        case 1
22          // Greedy algorithm
23          // ‘Approximation algorithms for combinatorial problems’ (1973)
24           $hs \leftarrow \text{FindGreedyHittingSet}(\mathcal{H})$ 
25        end
26       $(sat?, \mu) \leftarrow \text{SatSolver}(hs)$ 
27      if not sat? then
28        switch nonOptLevel do
29          case 0
30             $nonOptLevel \leftarrow 1$ 
31          case 1
32            break
33          end
34        else
35           $\mathcal{C} \leftarrow \mathcal{F} \setminus \text{Grow}(hs)$ 
36           $\mathcal{H} \leftarrow \mathcal{H} \cup \{\mathcal{C}\}$ 
37           $nonOptLevel \leftarrow 0$ 
38        end
39      end
40    end
41  end
42  return  $(hs', cost(hs))$ 
43 end
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### 3 MIP hitting set problem specification

For the set of clauses  $\mathcal{C} = \{c_1, \dots, c_{|\mathcal{C}|}\}$  with weights  $\mathcal{W} = \{w_1, \dots, w_{|\mathcal{C}|}\}$  in the collection of sets  $\mathcal{H}$ . For Example:

$$\begin{aligned}\mathcal{C} &= \{c_1, \dots, c_6\} \\ \mathcal{W} &= \{w_1 = 20, w_2 = 20, w_3 = 10, w_4 = 10, w_5 = 10, w_6 = 20\} \\ \mathcal{H} &= \{c_3\}, \{c_2, c_4\}, \{c_1, c_4\}, \{c_1, c_5, c_6\}\end{aligned}\tag{1}$$

The optimal hitting set can be formulated as an integer linear program.

$$\min \sum_{i \in \{1..|\mathcal{C}|\}} w_i \cdot x_i \tag{2}$$

$$\sum_{i \in \{1..|\mathcal{C}|\}} x_i \cdot h_{ij} \geq 1, \forall j \in \{1..|\mathcal{H}|\} \tag{3}$$

$$x_i = \{0, 1\} \tag{4}$$

- $w_i$  is the input cost/weight associated with clause  $i$  in
- $x_i$  is a boolean decision variable if constraint/clause  $c_i$  is chosen or not.
- Equation 3,  $h_{ij}$  is a boolean input variable corresponding to if constraint/clause  $i$  is in set to hit  $j$ .

## 4 Future Work

## References