

# 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    $\mathcal{I}_{end} \leftarrow \text{propagate}(\mathcal{I}_0, \mathcal{T})$  // Goal state
3    $\mathcal{I} \leftarrow \mathcal{I}_0$  // Initial partial interpretation
4    $Seq \leftarrow \text{empty set}$  // explanation sequence
5   while  $\mathcal{I} \neq \mathcal{I}_{end}$  do
6     for  $i \in \mathcal{I}_{end} \setminus \mathcal{I}$  do
7        $X_i \leftarrow \text{OMUS}(\{\neg i\} \wedge \mathcal{I} \wedge \mathcal{S})$ 
8        $E_i \leftarrow \mathcal{I} \cap X_i$  // Facts used
9        $S_i \leftarrow \mathcal{T} \cap X_i$  // Constraint used
10       $\mathcal{N}_i \leftarrow \text{propagate}(E_i \wedge S_i)$  // Newly derived facts
11    end
12     $(E_{best}, S_{best}, N_{best}) \leftarrow (E_i, S_i, N_i)$  with lowest  $f(E_i, S_i, N_i)$ 
13    append  $(E_{best}, S_{best}, N_{best})$  to  $Seq$ 
14     $\mathcal{I} \leftarrow \mathcal{I} \cup \{N_{best}\}$ 
15  end
16 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    $mode \leftarrow mode\_greedy$ 
4   while true do
5     // Find a series of non-optimal solutions
6     while true do
7       switch  $nonOptLevel$  do
8         case  $mode\_incr$ 
9           // Add/Remove clause (choose clause appears most frequently in the
           // set of hitting sets so far)
10           $hs \leftarrow \text{FindIncrementalHittingSet}(H, C, hs)$ 
11        case  $mode\_greedy$ 
12          // Greedy algorithm
13          // 'Approximation algorithms for combinatorial problems' (1973)
14           $hs \leftarrow \text{FindGreedyHittingSet}(\mathcal{H})$ 
15        end
16         $(sat?, \mu) \leftarrow \text{SatSolver}(hs)$ 
17        if not  $sat?$  then
18          switch  $nonOptLevel$  do
19            case  $mode\_incr$ 
20               $mode \leftarrow mode\_greedy$ 
21            case  $mode\_greedy$ 
22               $mode \leftarrow mode\_opt$ 
23              break
24            end
25          else
26             $C \leftarrow \mathcal{F} \setminus \text{Grow}(hs)$ 
27             $\mathcal{H} \leftarrow \mathcal{H} \cup \{C\}$ 
28             $mode \leftarrow mode\_incr$ 
29          end
30           $hs \leftarrow \text{OptimalHittingSet}(\mathcal{H}, cost)$  // Find optimal solution
31           $(sat?, \mu) \leftarrow \text{SatSolver}(hs)$ 
32          // If SAT,  $\mu$  contains the satisfying truth assignment
33          // IF UNSAT,  $hs$  is the OMUS
34          if not  $sat?$  then
35            break
36          end
37           $C \leftarrow \mathcal{F} \setminus \text{Grow}(hs)$ 
38           $\mathcal{H} \leftarrow \mathcal{H} \cup \{C\}$ 
39        end
40        return  $(hs', cost(hs))$ 
41      end
42    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