1 Explanation Generation

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

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 Γ be a collection of sets and $HS(\Gamma)$ the set of all hitting sets on Γ 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 Γ 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 $K \subseteq MCSes(\mathcal{F})$. Then a subset \mathcal{U} of \mathcal{F} is an OMUS if \mathcal{U} is a optimal hitting set on K and \mathcal{U} is unsatisfiable

```
Algorithm 2: OMUS-Delayed (\mathcal{F}, [f, \mathcal{H}_0])
    input
                            : \mathcal{F} a CNF formula
    input
                            : cost a cost function
    optional input: \mathcal{H}_0 initial collection of disjoint Minimum Correction Sets
    output
                            : \mathcal{OMUS}(\mathcal{F})
 1 begin
         \mathcal{H} \leftarrow \mathtt{DisjointMCS}(\mathcal{F})
 2
         while true do
 3
              hs \leftarrow \texttt{OptimalHittingSet}(\mathcal{H}, cost)
                                                                                                               // Find optimal solution
 4
              \mathcal{F}' \leftarrow \{c_i | e_i \in hs\}
 5
              (\text{sat?}, \kappa) \leftarrow \texttt{SatSolver}(\mathcal{F}')
 6
              // If SAT, \kappa contains the satisfying truth assignment
              // IF UNSAT, \mathcal{F}' is the OMUS
              if not sat? then
 7
                   break
 8
                   \mathcal{C} \leftarrow \mathcal{F} \setminus \mathtt{Grow}(\mathcal{F}')
 9
                   \mathcal{H} \leftarrow \mathcal{H} \cup \{\mathcal{C}\}
10
                   nonOptLevel \leftarrow 0
11
                    // Find a series of non-optimal solutions
                   while true do
12
                        switch nonOptLevel do
13
                              case \theta
14
                                   // adds clause to current core to the current hitting set
                                  hs \leftarrow \texttt{FindIncrementalHittingSet}(H, \mathcal{F}', hs)
15
                              case 1
16
                                  hs \leftarrow \texttt{FindGreedyHittingSet}(\mathcal{H})
17
                         end
18
                        \mathcal{F}' \leftarrow \{c_i | e_i \in hs\}
19
                        (\text{sat}?, \kappa) \leftarrow \texttt{SatSolver}(\mathcal{F}')
20
                        if not sat? then
21
                              switch nonOptLevel do
22
                                  case \theta
23
                                    nonOptLevel \leftarrow 1
24
                                   case 1
25
                                       break
26
                             end
27
                        else
28
                              \mathcal{C} \leftarrow \mathcal{F} \, \setminus \, \mathtt{Grow}(\mathcal{F}')
29
                              \mathcal{H} \leftarrow \mathcal{H} \cup \{\mathcal{C}\}
30
                              nonOptLevel \leftarrow 0
31
                   end
32
              end
33
34
         return (\mathcal{F}', cost(\mathcal{F}'))
36 end
```

3 MIP hitting set problem specification

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

$$C = \{c_1, ... c_6\}$$

$$W = \{w_1 = 20, w_2 = 20, w_3 = 10, w_4 = 10, w_5 = 10, w_6 = 20\}$$

$$\mathcal{H} = \{c_3\}, \{c_2, c_4\}, \{c_1, c_4\}, \{c_1, c_5, c_6\}$$

$$(1)$$

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

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

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

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

- \bullet 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