### 1 Formalization

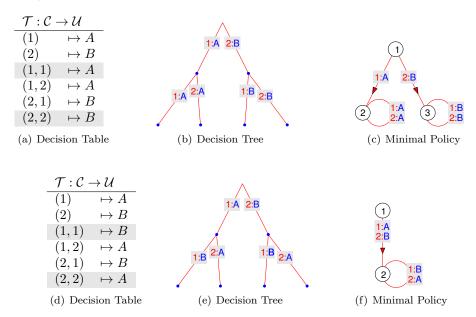
Let  $\mathcal{Y}$  be a set of observations and  $\mathcal{U}$  a set of control actions. The original formalization defines a decision table as a map  $\mathcal{T}: \mathcal{C} \to \mathcal{U}$ , where  $\mathcal{C} = \bigcup_{i=1}^n \mathcal{Y}^i$  for some  $n \in \mathbb{N}$ . This definition is often stronger than necessary, as there may be contexts c in which certain observations  $y \in \mathcal{Y}$  are never encountered. In such a context, any assignment  $\mathcal{T}(c) \in \mathcal{U}$ , produces the same policy, although the choice of  $\mathcal{T}(c)$  can affect a policy's reducibility.

Example 1 (Equivalent Decision Tables) Suppose  $\mathcal{Y} = \{1, 2\}$ ,  $\mathcal{U} = \{A, B\}$ ,  $\mathcal{C} = \bigcup_{i=1}^{2} \mathcal{Y}^{i}$  and

$$\mathcal{T}(c) = \begin{cases} A & c \in \{(1), (1, 2)\} \\ B & c \in \{(2), (2, 1)\} \end{cases},$$

$$\mathcal{T}'(c) & otherwise.$$

is an optimal policy, for arbitrary  $\mathcal{T}': \mathcal{C} \to \mathcal{U}$ . However, depending on the choice of  $\mathcal{T}'$  (highlighted in the tables below), the completed policy can have differently-sized minimal representations:



Instead, we propose an "incompletely-determined" formalization:

**Definition 1 (Policies)** Given a set  $\mathcal{Y}$  of observations, recursively construct

$$C_0 = \{\emptyset\} \quad and \quad C_{i+1} = \{(c, y_{i+1}) : c \in C_i, y_{i+1} \in \mathcal{Y}_c\},$$
 (1)

where  $\mathcal{Y}_c \subseteq \mathcal{Y}$  are the observations that may be seen in context c. Let  $\mathcal{C} = \bigcup_{i=0}^{\infty} \mathcal{C}_i$ . A **policy** P is then a tuple  $\langle \mathcal{C}, \mathcal{U}, \mathcal{T}, \mathcal{Y} \rangle$ , where  $\mathcal{U}$  is some decision set and  $\mathcal{T} : \mathcal{C} \setminus \{\emptyset\} \to \mathcal{U}$ .

**Definition 2 (Completely-Determined Policies)** If  $\mathcal{Y} = \bigcup_{\mathcal{C}} \mathcal{Y}_c$  and  $\mathcal{C} = \mathcal{Y}^{\leq n}$  for some  $n \in \mathbb{N}$ , then  $P = \langle \mathcal{C}, \mathcal{U}, \mathcal{T}, \mathcal{Y} \rangle$  is completely determined. A completion of P is a policy  $P' = \langle \mathcal{C}', \mathcal{U}', \mathcal{T}', \mathcal{Y} \rangle$  such that

$$\mathcal{Y} \subseteq \mathcal{Y}', \quad \mathcal{C}' = \bigcup_{i=0}^{\infty} (\mathcal{Y}')^i, \quad \mathcal{U} \subseteq \mathcal{U}', \quad and \quad \mathcal{T}'|_{\mathcal{C}} = \mathcal{T}.$$
 (2)

Let Comp(P) be the set of completions of the policy P.

**Definition 3 (FSM Representations)** An **FSM representation** (or just **representation**) is a tuple  $\langle \mathcal{C}, \mathcal{R}, \mathcal{U}, \mathcal{S}, \mathcal{T}, \mathcal{Y} \rangle$  (abbreviated to  $\langle \mathcal{R}, \mathcal{S} \rangle$  when  $P = \langle \mathcal{C}, \mathcal{U}, \mathcal{T}, \mathcal{Y} \rangle$  is given), with "states"  $\mathcal{S} \subseteq \mathbb{N}$  and state assignments  $\mathcal{R} : \mathcal{C} \to \mathcal{S}$ , such that

$$\mathcal{R}(c) = \mathcal{R}(c')$$
 and  $y \in \mathcal{Y}_c \cap \mathcal{Y}_{c'} \implies \mathcal{T}(c, y) = \mathcal{T}(c', y)$ . (3)

Let Rep(P) be the set of representations of the policy P.

**Definition 4 (Minimal Representations)** The size of an FSM representation is the cardinality of its state set. A representation  $\langle \mathcal{R}, \mathcal{S} \rangle$  of P is **minimal** if  $|\mathcal{S}| = \min\{|\mathcal{S}'| : \langle \mathcal{R}', \mathcal{S}' \rangle \in \operatorname{Rep}(P)\}$ . A representation  $\langle \mathcal{R}', \mathcal{S}' \rangle$  is a **reduction** of the representation  $\langle \mathcal{R}, \mathcal{S} \rangle$  if there is a surjection  $\phi : \mathcal{S} \to \mathcal{S}'$  such that  $\mathcal{R}' = \phi(\mathcal{R})$ .

**Example 2** If  $C = \{c_1, c_2, \ldots\}$ , then we have a canonical representation  $(\mathcal{R}, \mathcal{S})$ , where

$$S = \{1, \dots, |\mathcal{C}|\} \quad and \quad \mathcal{R} : c_k \mapsto k. \tag{4}$$

It can be shown that the size of a minimal representation of a policy P is equal to the minimum size of the minimal representations of its completions, i.e.

$$\min\{|\mathcal{S}'| : \langle \mathcal{R}', \mathcal{S}' \rangle \in \operatorname{Rep}(P)\} = \min\{|\mathcal{S}'| : \langle \mathcal{R}', \mathcal{S}' \rangle \in \operatorname{Rep}(P'), P' \in \operatorname{Comp}(P)\}$$
 (5)

Incompletely-determined policies allow more freedom in representation reduction, as shown in the next example.

Example 3 (Incompletely-Determined Policies) Let  $C = \{\emptyset, (1), (2), (1, 2), (2, 1)\}, \ \mathcal{U} = \{A, B\},$ and

$$\mathcal{T}(c,y) = \begin{cases} A & c \in \{(1), (1,2)\} \\ B & c \in \{(2), (2,1)\} \end{cases}.$$

Observe that the minimal policy is the same as that of the completely-determined policy in Example 1(f).

# 2 Algorithm

To find a minimum representation of a given policy, we first compute a graph of reducibility relations, then compute a minimal clique-covering. For practical computation of reducibility, we'll start with the weaker condition of compatibility.

#### 2.1 Reducibility Relations

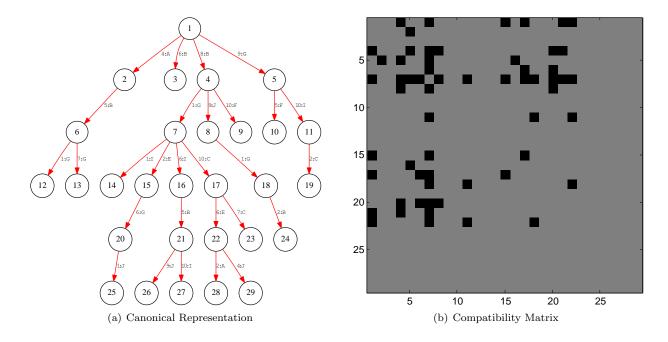
**Definition 5 (Reducibility)** For a given policy  $P = \langle \mathcal{C}, \mathcal{U}, \mathcal{T}, \mathcal{Y} \rangle$ , two contexts  $c_1, c_2 \in \mathcal{C}$  are **reducible** (write  $c_1 \sim c_2$ ) if there exists a representation  $\langle \mathcal{R}, \mathcal{S} \rangle$  of P such that  $\mathcal{R}(c_1) = \mathcal{R}(c_2)$ . Likewise, for a given representation  $R = \langle \mathcal{R}, \mathcal{S} \rangle$ , two states  $s_1, s_2 \in \mathcal{S}$  are **reducible** if there exists a reduction  $(\phi, \langle \mathcal{R}', \mathcal{S}' \rangle)$  of R such that  $\phi(s_1) = \phi(s_2)$ .

Observe that for any representation  $\langle \mathcal{C}, \mathcal{R}, \mathcal{U}, \mathcal{S}, \mathcal{T}, \mathcal{Y} \rangle$ , the contexts  $c_1, c_2 \in \mathcal{C}$  are reducible if and only if the states  $\mathcal{R}(c_1)$  and  $\mathcal{R}(c_2)$  are reducible. Observe also that for incompletely-determined policies, reducibility is a symmetric but not-necessarily-transitive relation

**Example 4 (Non-Transitive Reducibility)** Suppose  $\mathcal{Y} = \{1, 2, 3\}$ ,  $\mathcal{C} = \{\emptyset, (1), (2), (1, 3), (2, 3)\}$ ,  $\mathcal{U} = \{A, B\}$ , and

$$\mathcal{T}(c) = \begin{cases} A & c \in \{(1), (1,3)\} \\ B & c \in \{(2), (2,3)\} \end{cases}$$
 (6)

Observe that, under this policy,  $\emptyset \sim (1)$  and  $\emptyset \sim (2)$ , but  $(1) \not\sim (2)$ , since  $\mathcal{T}(1,3) \neq \mathcal{T}(2,3)$ .



However, it can be shown that, under a completely-determined policy, reducibility induces an equivalence relation. In either case, we compute reducibility using the following criterion:

**Lemma 1** Two contexts  $c_1, c_2 \in \mathcal{C}$  are reducible iff

**Output:** A reducibility matrix  $A: \mathcal{S} \times \mathcal{S} \to \{\text{true}, \text{false}\}.$ 

$$\mathcal{T}(c_1, s) = \mathcal{T}(c_2, s) \quad \text{for all} \quad s \in \mathcal{Y}^* \quad \text{such that} \quad (c_1, s), (c_2, s) \in \mathcal{C}$$
 (7)

This informs the following algorithm

**Input:** A representation  $\langle \mathcal{C}, \mathcal{R}, \mathcal{U}, \mathcal{S}, \mathcal{T}, \mathcal{Y} \rangle$ 

end for end if end for until  $\sim$  is Changed

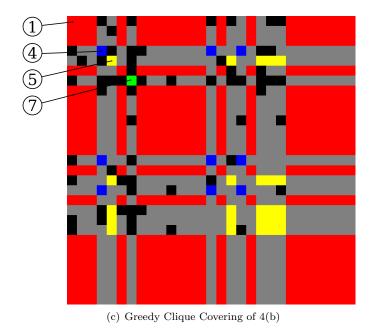
#### Algorithm 1: Compute Reducibility Relations

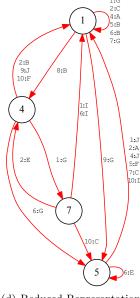
```
A(s_1,s_2) \Leftarrow \mathbf{true} \quad \text{for all } s_1,s_2 \in \mathcal{S}. repeat isChanged \Leftarrow \mathbf{false} for s_1 < s_2 \in \mathcal{S} do \mathbf{if } A(s_1,s_2) = \mathbf{true \ then} for c_1 \in \mathcal{R}^{-1}(s_1), \, c_2 \in \mathcal{R}^{-1}(s_2) do \mathbf{for } y \in \mathcal{Y}_{c_1} \cap \mathcal{Y}_{c_2} \mathbf{do} \mathbf{if } \mathcal{T}(c_1,y) \neq \mathcal{T}(c_2,y) \text{ or } ^{\sim}A(\mathcal{R}(c_1,y),\mathcal{R}(c_2,y)) \mathbf{then} A(s_1,s_2) \Leftarrow \mathbf{false}. isChanged \Leftarrow \mathbf{true}. end if end for
```

Now, although reducibility is not an equivalence relation, any reduction  $\phi: \mathcal{S} \to \mathcal{S}'$  induces an equivalence relation, partitioning  $\mathcal{S}$  into cliques of mutually-reducible states, i.e.

$$S = \bigsqcup_{s' \in S'} \phi^{-1}(s'), \quad \text{where} \quad \phi(s_1) = \phi(s_2) \implies A(s_1, s_2)$$
 (8)

Thus, a minimal reduction induces a minimal clique partition of the reducible states of a representation.





(d) Reduced Representation

### 2.2 Assembling Cliques

end while

Notation 1 (Arrow notation) For a policy  $P = \langle \mathcal{C}, \mathcal{U}, \mathcal{T}, \mathcal{Y} \rangle$ , write  $c \to c'$  if  $c = (c_1, \ldots, c_i) \in \mathcal{C}_i \subseteq \mathcal{C}$  and  $c' = (c_1, \ldots, c_i, y) \in \mathcal{C}_{i+1} \subseteq \mathcal{C}$ , for some i. For a representation  $\langle \mathcal{R}, \mathcal{S} \rangle$  of P, write  $s_1 \to s_2$  if there are  $c_1 \in f^{-1}(s_1)$  and  $c_2 \in f^{-1}(s_2)$  such that  $c_1 \to c_2$ .

We propose the following, greedy, approximate algorithm Although we have no proof that this algorithm

```
Algorithm 2: Compute Clique Covering
```

```
Input: A representation \langle \mathcal{C}, \mathcal{R}, \mathcal{U}, \mathcal{S}, \mathcal{T}, \mathcal{Y} \rangle with s_1 < s_2 only if s_2 \not\rightarrow s_1.

Input: A reducibility matrix A: \mathcal{S} \times \mathcal{S} \to \{\mathbf{true}, \mathbf{false}\} as computed by Algorithm 1.

Output: A partition function \phi: \mathcal{S} \to \mathcal{S}' with \phi(s_1) = \phi(s_2) only if A(s_1, s_2).

\mathcal{S}' \Leftarrow \mathcal{S}
\phi \Leftarrow id_{\mathcal{S}}
unused \Leftarrow \mathcal{S}
while |unused| > 0 do
s_1 \Leftarrow \min(unused)
unused \Leftarrow unused \setminus \{s_1\}
for s_2 \in unused do
if A(s_1, s_2) then
\phi(s_2) \Leftarrow s_1
unused \Leftarrow unused \setminus \{s_2\}
end if
end for
```

produces minimal representations of a given policy, it is not inconceivable that this or another greedy policy could work. In general, the Minimal Clique Covering problem is NP-Complete, but the tree structure of the decision policy is an a constraint that may simplify the problem.

## 3 Code

The main MATLAB routine is in decision\_script.m. By default it generates a random decision tree, whose branching is governed by a Poisson distribution.

### 3.1 Console Output

#### Initial FSM:

```
01: 08-->(05, 02),
     02-->(05, 03),
                      06 -> (03, 04), 10 -> (01, 05),
     07 --> (04, 06),
                      08-->(07, 07),
03:
04:
                      07 --> (06, 09),
05:
     05-->(10, 08),
06:
     04 --> (09, 10),
                      05 --> (05, 11), 07 --> (09, 12),
07:
08:
     09 --> (07, 13),
     03-->(08, 14),
09:
10:
11:
     04-->(01, 15),
12:
                      06 --> (06, 16),
13:
     01-->(03, 17),
     07 --> (01, 18),
14:
15:
16:
17:
18:
```

Final clique covering: {1, 2, 4, 5, 7, 8, 9, 10, 11, 13, 15, 16, 17, 18} {3, 12} {6} {14}

#### 3.2 Visualizations

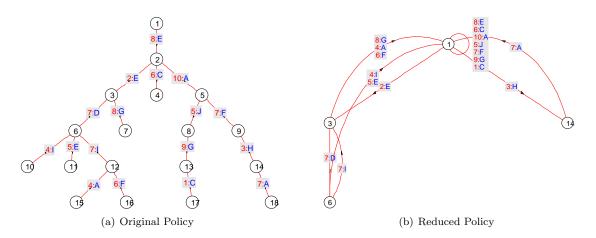


Figure 4: MATLAB Visualizations

An optional section of this script uses Graphviz to produce high-quality FSM visualizations.

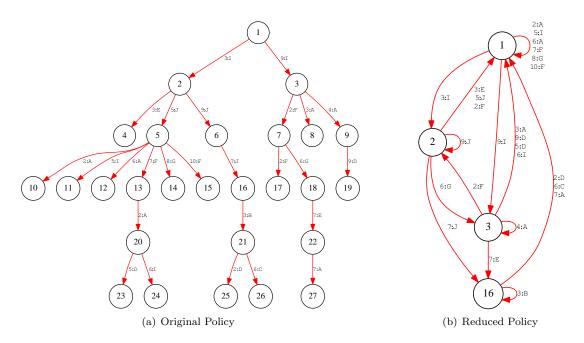


Figure 5: Graphviz Visualizations

This makes a system call to the script <code>drawgraph.sh</code>, which runs in the bash shell, and uses a few non-standard graphics utilities. In Ubuntu, these dependencies should be resolved by the command

\$ sudo apt-get install ghostscript texlive-extra-utils graphviz evince