

Lecture 11: Learning a DFA

October 2, 2019

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11.1 Review

11.1.1 Definition

Definition 11.1 A *Membership query* consists of a string $w \in \Sigma^*$. The answer is either yes or no depending on whether w is in this language.

Definition 11.2 *Equivalence queries* take a hypothesis DFA. If the hypothesis DFA is identical to M , the output is yes. Otherwise the output is no and a counterexample (A counterexample is a string that is accepted by the hypothesis DFA but not by M or vice versa).

11.1.2 Observation Table

The Angluin's Algorithm maintains a **observation table** which holds the set of candidate states of a DFA S and a set of query strings E . The observation table has two properties as follows.

- Rows are labeled by strings, which are candidate states. And they are prefix closed.
- Columns are query strings, which are suffix closed.

There are two attributes of the observation table.

- **Closed:** $row(s) \in S \implies \forall a \in \Sigma, row(s \cdot a) \in S$.
- **Consistent:** $\forall s_1, s_2, row(s_1) = row(s_2) \implies \forall a \in \Sigma, row(s_1 \cdot a) = row(s_2 \cdot a)$.

The rule for the entries of observation table is defined as:

$$T(s, e) = \begin{cases} 1 & \text{for } s \cdot e \in L \\ 0 & \text{for } s \cdot e \notin L \end{cases}$$

The interpretation of $T(s, e)$ is that $T(s, e)$ is 1 if and only if s concatenated by e is a string in L .

11.2 Angluin's Algorithm

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1 Start with  $S = E = \{\varepsilon\}$  and  $T(\varepsilon), T(0), T(1)$  are given.
2 if the observation table is not consistent, i.e.,  $\exists a \in \Sigma$ , s.t.  $row(s_1) = row(s_2)$ , but
    $row(s_1 \cdot a) \neq row(s_2 \cdot a)$ , i.e.,  $\exists e \in E$ , s.t.,  $T(s_1 \cdot a \cdot e) \neq T(s_2 \cdot a \cdot e)$  then
3   | add  $a \cdot e$  and all its suffixes to  $E$ , complete the observation table
4 if the observation table is not closed, i.e.,  $\exists t \in S \cdot E$ , s.t.  $row(t) \neq row(s), \forall s \in S$  then
5   | add  $t$  and all its prefixes to  $S$ , complete the observation table
6 Repeat 2 – 5 until the observation table is consistent and closed.
7 if the observation table is consistent and closed then
8   | proposed a DFA based on it if there is an counterexample  $x$  then
9     | add  $x$  and all its prefixes to  $S$ , complete table.
10    | Repeat 2 – 7
11   else
12    | The proposed DFA is what we want.

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Given a closed and consistent observation table, we define a corresponding DFA over alphabet Σ with set of states Q , initial state q_0 , set of final states F and transition function δ as followed:

- $Q = \{row(s) : s \in S\}$,
- $q_0 = row(\varepsilon)$,
- $F = \{row(s) : s \in S \text{ and } T(s) = 1\}$
- $\delta(row(s), a) = row(s \cdot a)$

For new DFA M we just constructed, each state represents a unique row in observation matrix. This DFA is a feasible DFA. Furthermore, if other DFAs that is acceptance consistent with M but is not identical to M must have more states than M .

Lemma 11.3 *DFA $M(T)$ is consistent with table T and M is the smallest such DFA.*

Proof: To prove the lemma, we have to prove two things:

1. $\delta(q_0, s) = row(s)$
2. $\delta(q_0, s \cdot e) \in F, \iff T(s \cdot e) = 1$

First, we prove the first part. We conduct induction on $|s|$.

Base: $\delta(q_0, \varepsilon) = row(\varepsilon)$ holds.

We assume 1. holds for $|s| = k$.

For $|s| = k + 1$, $\delta(q_0, s) = \delta(q_0, s_1 \cdot a) = \delta(\delta(q_0, s_1), a) = \delta(row(s_1), a) = row(s_1 \cdot a) = row(s), \exists a \in \Sigma, s = s_1 \cdot a$.

Hence, the first part is correct.

We can do the similar induction for the second part and we leave this part for you to complete. ■

From 11.3, we know that any DFA consistent with table T must have distinct states corresponding to distinct rows of S .

Lemma 11.4 *At most $n - 1$ conjectures including not closed or not consistent.*

Let's consider the time complexity of Angluin's Algorithm. Given n states, alphabet size k and the longest counterexample m , the time complexity is linear with $m \cdot n \cdot k$, i.e., $poly(n, m, k)$.