Contents lists available at ScienceDirect

## Systems & Control Letters

journal homepage: www.elsevier.com/locate/sysconle

# Trajectory detectability of discrete-event systems

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#### ARTICLE INFO

Article history: Received 19 June 2017 Received in revised form 14 July 2018 Accepted 16 July 2018

Keywords: Discrete-event systems Trajectory estimation Detectability

### ABSTRACT

We investigate the trajectory estimation problem for partially-observed discrete event systems. In some applications, only knowing the current state of the system may be insufficient, and knowing which trajectory the system takes to reach the current state could be important. This requires more precise knowledge about the system. In this paper, a language-based framework is proposed in order to tackle this problem. Two new notions of detectability, called trajectory detectability and periodic trajectory detectability, are proposed to capture different requirements in the aforementioned trajectory estimation problem. Effective verification algorithms are also provided. Our results extend the theory on detectability of discrete event systems from state estimation problem to trajectory estimation problem.

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#### 1. Introduction

System estimation is one of the central problems in systems and control theory. In many applications, we do not have full access to the system's internal state and need to perform state estimation. The state estimation problem becomes particularly important when one wants to make decisions based on the limited system information. In this paper, we investigate the state estimation problem for partially-observed Discrete-Event Systems (DES).

The problem of state estimation has drawn considerable attentions in the DES literature due to its importance; see, e.g., [1–6]. This problem was initiated in [1,2], where the notion of observability was defined.<sup>1</sup> Recently, the state estimation problem has been studied more systematically in the framework of detectability; see, e.g., [4,5,8–14]. Particularly, in [4], the authors define four types of detectability in order to capture different requirements in practice. These notions of detectability have been further generalized by [5,8]. For example, [8] defines a generalized detectability based on the state disambiguation problem [3,6]. Detectability has also been studied in the framework of stochastic DES by [9,11]. When the original system is not detectable, several approaches have been proposed in order to enforce detectability, e.g., by sensor activations [15,16] and by supervisory control [17,18]. State

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<sup>1</sup> The notion of observability was also used in [7] as a condition under which a partial-observation supervisor can correctly make control decisions.

https://doi.org/10.1016/j.sysconle.2018.07.008 0167-6911/© 2018 Elsevier B.V. All rights reserved. estimation problem has also been investigated in the context of colored graph [19].

All of the aforementioned works on detectability are *state-based*. Namely, one wants to estimate the current state of the system based on a given model. However, in some applications, knowing the current state of the system is not sufficient. For example, in the application of location-based services (LBS) [20], a DES is usually used to represent the connective of a region and each state in it corresponds to a location. Sometimes, however, simply knowing the current location may not be sufficient, for instance, if we want to know by which path this location is reached. Therefore, instead of estimating the state of the system, one may also be interested in estimating the *trajectory* of the system.

In this paper, we systematically study the trajectory estimation problem in the context of partially-observed DES. Specifically, this paper has the following contributions. First, we define the notions of trajectory detectability and periodic trajectory detectability. These two notions provide the conditions for determining *a priori* if the trajectory of a given system can be determined after a bounded delay or be determined periodically. Second, for regular languages, i.e., languages that can be marked by finite-state automata, we provide effective algorithms to verify these two conditions. In particular, the verification algorithm for trajectory detectability requires polynomial-times using a twin-machine-like construction. On the other hand, the algorithm for verifying periodic trajectory detectability requires exponential complexity.

Note that, although the study of trajectory detectability is motivated by state detectability [4], their verifications are quite different. In general, we can always refine an automaton by expanding







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its state space such that each state carries more information. However, the domain of language is infinite and one may not always be able to use finite states to precisely capture the (infinite) trajectory information. Our framework is fully language-based and it does not depend on the automaton realizing the language. Moreover, we show that infinite language information can still be effectively verified by using its underlying automaton. In particular, we only exploit the standard twin-machine construction [21,22], which is different from the detector construction proposed in [8]. We show that the phenomenon of *information merge* plays an important role in the trajectory estimation problem. This issue also does not exist in the state-based framework for detectability.

We also would like to remark that our paper is not the first one investigating the trajectory estimation problem in DES. In [23], the authors investigate a similar problem and the notion of *invertibility* is proposed. Our work is different from [23] due to the following reasons. First, we systematically investigate both trajectory detectability and periodic trajectory detectability; both of these two notions are different from invertibility. Specially, invertibility only requires to recover the last *n* events but detectability requires to recover the precise trajectory executed by the system. Second, we provide a language-based framework for studying this problem, while the result in [23] is state dependent. Finally, invertibility is defined only for prefix-closed languages, while trajectory detectability is defined for non-prefix-closed languages.

#### 2. Preliminaries

Let  $\Sigma$  be a finite set of events. A string  $s = \sigma_1 \dots \sigma_n$  is finite sequence of events and we denote by |s| the length of s. We use  $\epsilon$  to denote the empty string with  $|\epsilon| = 0$ . We denote by  $\Sigma^*$ the set of all strings including  $\epsilon$ . A language  $L \subseteq \Sigma^*$  is a set of strings. The prefix-closure of language L is defined as  $\overline{L} := \{w \in \Sigma^* : \exists v \in \Sigma^* \text{ s.t. } wv \in L\}$ . We say that L is prefix-closed if  $\overline{L} = L$ . We denote by L/s the post-language of L after string s, i.e.,  $L/s := \{t \in \Sigma^* : st \in L\}$ . We denote by Card[L] the cardinality of L, which is the number of strings in L. For two strings  $s, t \in \Sigma^*$ , we write  $s \leq t$  if  $s \in \overline{\{t\}}$ .

A deterministic finite-state automaton (DFA) is a 5-tuple  $G = (Q, \Sigma, \delta, q_0, Q_m)$ , where Q is the finite set of states,  $\Sigma$  is the finite set of events,  $q_0$  is the initial state,  $Q_m$  is the set of marked states and  $\delta : Q \times \Sigma \rightarrow Q$  is the partial transition function, where  $\delta(q, \sigma) = q'$  means that there exists a transition labeled with event  $\sigma$  from state q to state q'. The transition function is also extended to  $Q \times \Sigma^*$  in the usual manner; see, e.g., [24]. We denote by  $\mathcal{L}(G)$  the language generated by G, i.e.,  $\mathcal{L}(G) = \{s \in \Sigma^* : \delta(q_0, s)!\}$ , where ! means "is defined". We denoted by  $\mathcal{L}_m(G)$  the language marked by G, i.e.,  $\mathcal{L}(G) = \{s \in \Sigma^* : \delta(q_0, s)!\}$ , where !  $C \in \Sigma^*$  is regular if there exists a DFA G such that  $\mathcal{L}_m(G) = L$ .

In many cases, the event generated by the system cannot be observed perfectly. Therefore, we assume that the event set  $\Sigma$  is partitioned into two disjoint sets  $\Sigma = \Sigma_o \dot{\cup} \Sigma_{uo}$ , where  $\Sigma_o$  is the set of observable events and  $\Sigma_{uo}$  denotes the set of unobservable events. Then  $P : \Sigma^* \to \Sigma_o^*$  denotes the natural projection that erases event in  $\Sigma_{uo}$  from a string; this can be defined by

$$P(\epsilon) = \epsilon \text{ and } P(s\sigma) = \begin{cases} P(s)\sigma & \text{if } \sigma \in \Sigma_o \\ P(s) & \text{if } \sigma \in \Sigma_{uo} \end{cases}$$
(1)

The natural projection is also extended to  $2^{\Sigma^*}$  by  $P(L) = \{s \in \Sigma_o^* : \exists t \in L \text{ s.t. } P(t) = s\}$ . We denote by  $P^{-1}$  the inverse projection.

Given a DFA *G* and a set of states  $I \subseteq Q$ , we denote by  $Acc_G(I)$  the set of states accessible from some state in *I*, i.e.,

$$\operatorname{Acc}_{G}(I) = \{ q \in Q : \exists q' \in I, \exists s \in \Sigma^{*} \text{ s.t. } \delta(q', s) = q \}.$$

$$(2)$$



**Fig. 1.** For both  $G_1$  and  $G_2$ :  $\Sigma_o = \{o\}$  and  $\Sigma_{uo} = \{a, b\}$ .

Let  $q \in Q$  be a state in *G*. We denote by  $In_G(q)$  the set of events entering *q*, i.e.,

$$In_{G}(q) = \{ \sigma \in \Sigma : \exists q' \in Q \text{ s.t. } \delta(q', \sigma) = q \}.$$
(3)

Finally, let  $s \in P(\mathcal{L}(G))$ , we denote by  $R_G(s)$  the set of states that can be reached by observing *s*, i.e.,

$$R_{\mathcal{G}}(s) = \{q \in Q : \exists t \in \Sigma^* \text{ s.t. } \delta(q_0, t) = q \land P(t) = s\}.$$
(4)

#### 3. State-Based detectability and trajectory-based detectability

Due to measurement uncertainty, one may not always have a perfect knowledge about the current status of the system. In [4], the notion of (strongly) detectability was introduced in order to capture whether or not we can eventually have a perfect knowledge about the system after finite delay. In this paper, we refer to detectability defined in [4] as *state detectability*. First, we recall its definition.

**Definition 1.** A DFA  $G = (Q, \Sigma, \delta, q_0, Q_m)$  is said to be *state* detectable w.r.t.  $\Sigma_o$  if

$$(\exists n \in \mathbb{N})(\forall s \in L)[|P(s)| \ge n \Rightarrow |R_G(P(s))| = 1]$$
(5)

**Example 1.** Let us consider system  $G_1$  shown in Fig. 1(a), where  $\Sigma_o = \{o\}$ . Clearly, this system is state detectable since  $R_G(o^n) = \{5\}$  for any  $n \ge 2$ . However, system  $G_2$  shown in Fig. 1(b) is not state detectable. To see this, for any  $n \ge 2$ , we can find  $o^n$  such that  $R_G(o^n) = \{5, 6\}$ , i.e., we can never determine the current state of the system precisely.

Intuitively, state detectability says that, after a finite delay, we will know exactly the current state of the system and maintain this ability in the future. However, in some applications, this requirement may be too strong. Therefore, in [4], the notion of periodic state detectability was also proposed, which only requires that we can detect the state of the system periodically.

**Definition 2.** A DFA  $G = (Q, \Sigma, \delta, q_0, Q_m)$  is said to be *periodically* state detectable w.r.t.  $\Sigma_o$  if

$$(\exists n \in \mathbb{N})(\forall s \in L)(\forall t \in L/s : |P(t)| \ge n)$$
  
$$(\exists t' \le t)[|R_G(P(st'))| = 1]$$
(6)

**Example 2.** Let us consider system  $G_3$  shown in Fig. 1(c) with  $\Sigma_o = \{o\}$ . This system is not state detectable, since we cannot distinguish states 4 and 5 after observing  $oo(ooo)^n$  for any  $n \ge 0$ . However, it is periodically state detectable, since we always know for sure that the current state is 0 after observing  $(ooo)^n$  for any  $n \ge 0$ .

**Remark 1.** Note that the system automaton *G* considered in [4] has multiple initial states; say  $Q_0 \subseteq Q$ . In the above definitions, we only consider the case where the initial state is unique. However,

this setting is just for the sake of simplicity and is without loss of generality, since we can add a new single initial state and add unobservable transitions from this new initial state to each state in  $Q_0$ , which essentially simulates the case of multiple initial states.

In Example 1, we see that, although  $G_1$  is state detectable but  $G_2$  is not, they generate the same language, i.e.,  $\mathcal{L}(G_1) = \mathcal{L}(G_2)$ . In other words, a system, which is originally state detectable, may become non-detectable after refining its state-space and vice versa. In fact, a state in an automaton only tells the system's status about potential future behaviors. However, some information about the history is lost. In some cases, this history information is also very important. For example, in Fig. 1(a), even though we know for sure that the system is at state 5 after observing *oo*, we still cannot determine *how* state 5 is reached, since both *aoo* and *boo* are potential trajectories. Motivated by the above discussion, we define the notion of trajectory detectability.

**Definition 3.** A language *L* is said to be *trajectory detectable* w.r.t.  $\Sigma_0$  if

$$(\exists n \in \mathbb{N})(\forall s \in L)[|P(s)| \ge n \Rightarrow Card[P^{-1}P(s) \cap L] = 1]$$
(7)

**Definition 4.** A language *L* is said to be *periodically trajectory detectable* w.r.t.  $\Sigma_o$  if

$$(\exists n \in \mathbb{N})(\forall s \in L)(\forall t \in L/s : |P(t)| \ge n)$$
  
$$(\exists t' \le t)[Card[P^{-1}P(st') \cap L] = 1]$$
(8)

Intuitively, trajectory detectability requires that, after a finite delay, we are able to precisely recover the trajectory that leads to the current state and maintain this ability in the future. Similarly, periodic trajectory detectability requires that, we are able to precisely recover the trajectory of the system periodically.

Note that language *L* in the definition does not need to be prefixclosed. In fact, we consider non-prefix-closed language since it is more general than the prefix-closed case and using non-prefixclosed languages provides us more flexibility to describe the system's behavior of interest. This point is illustrated by the following example.

**Example 3.** Let us consider the DFA shown in Fig. 2(a) with  $\Sigma_o = \{o\}$ . Note that  $G_4$  is neither state detectable nor  $\mathcal{L}(G_4)$  is trajectory detectable. However, non-prefix-closed language  $\mathcal{L}_m(G_4)$  is trajectory detectable. Clearly, by observing  $o^{2k+1}$ ,  $k \leq 0$ , we know that the only possible string in  $\mathcal{L}_m(G)$  is  $ao^{2k+1}$ . Similarly, by observing  $o^{2k}$ ,  $k \leq 1$ , we know that the only possible string in  $\mathcal{L}_m(G_4)$  is  $bo^{2k+1}$ . This non-prefix-closed language can be used to model the case, where we are only interested in strings in  $\mathcal{L}_m(G)$  and do not care whether or not a wrong estimation is made for strings in  $\mathcal{L}(G) \setminus \mathcal{L}_m(G)$ . For instance, here we are only interested in distinguishing strings leading to state 2 and strings leading to state 3. Similarly, for  $G_5$  shown in Fig. 2(b), we know that  $\mathcal{L}_m(G_5)$  is periodically trajectory detectable but  $\mathcal{L}(G_5)$  is not. These examples justify our early assertion that using non-prefix-closed languages provides us more flexibility to describe the system's behavior of interest.

The following two results establish some relationships between state detectability and trajectory detectability. Their proofs are omitted, since the results follow directly from the definitions.

**Proposition 1.** Let *G* be a DFA and  $\Sigma_o$  be a set of observable events. If *G* is not (periodically) state detectable, then  $\mathcal{L}(G)$  is not (periodically) trajectory detectable.



**Fig. 2.**  $\mathcal{L}_m(G)$  is trajectory detectable but  $\mathcal{L}(G)$  is not trajectory detectable, where  $\Sigma_o = \{o\}$  and  $\Sigma_{uo} = \{a, b, c\}$ .

**Remark 2.** Note that the opposite direction of the above statement is not true in general; an example has been provided in Fig. 1(a). Moreover, when the language under consideration is non-prefix-closed, (periodic) trajectory detectability is no longer stronger than (periodic) state detectability in general; this issue has already been discussed in Example 3.

Given a DFA *G*, we say that *G* is a tree if  $\forall s, t \in \mathcal{L}(G) : s \neq t \Rightarrow \delta(q_0, s) \neq \delta(q_0, t)$ . The following result reveals that (periodic) state detectability and prefix-closed (periodic) trajectory detectability coincide when *G* is a tree.

**Proposition 2.** Let *G* be a tree and  $\Sigma_o$  be a set of observable events. Then *G* is (periodically) state detectable if and only if  $\mathcal{L}(G)$  is (periodically) trajectory detectable.

**Remark 3.** Intuitively, if *G* is a tree, then states and trajectories carry the same information. Clearly, if a language is finite, or equivalently, *G* is acyclic, then we can always refine *G* as a tree and make sure that each state in *G* uniquely carries the corresponding trajectory information. However, if we consider the general case, the following difficulties arise. First, we need to deal with non-prefix-closed languages. As we discuss earlier, trajectory detectability for the non-prefix-closed case is incomparable with either state detectability or its prefix-closed case. More importantly, in general, the language under consideration is infinite, i.e., *G* is not acyclic. Consequently, refining *G* as a tree or another structure requires infinite memory and it is not clear whether or not this trajectory information can be effectively verified. Therefore, we need new methods to handle these difficulties.

Remark 4. The definition of trajectory detectability is related to several language-based properties in the literature, e.g., opacity [25], diagnosability [26] and normality [7]. In [25], a security property called opacity was defined. Particularly, given two languages  $K, L \subset \mathcal{L}(G)$ , we say that (L, K) is weakly opaque if  $\exists s \in L, \exists t \in K : P(s) = P(t)$ . This definition is similar to the negation of trajectory detectability. However, they have the following main differences. First, weak opacity is a indistinguishable property that should hold for one specific string s. However, non-trajectory-detectability is a indistinguishable property that should hold eventually for all continuation strings starting from a sufficiently long point. Moreover, for weak opacity, language K to be distinguished from is static in the sense that it does not depend on the string considered in L. However, for trajectory detectability, for each string  $s \in \mathcal{L}_m(G)$ , we need to distinguish it from strings in  $\mathcal{L}_m(G) \setminus \{s\}$ . Therefore, the language to be distinguished from is "dynamic" in the sense that it depends on the string considered in  $\mathcal{L}_m(G)$ . Trajectory detectability is also related to diagnosability [26] since both of them require some information ambiguity after a finite delay. Particular, diagnosability requires to distinguish fault strings from non-fault strings after some delays, where the language to be distinguished from is still "static". This situation is the same for normality [7], where we still need to distinguish two "static" languages.

#### 4. Verification of trajectory detectability for regular languages

In this section, we investigate the verification of trajectory detectability for the case where the language *L* under consideration is *regular*. Therefore, we denote by  $G = (Q, \Sigma, \delta, q_0, Q_m)$  the DFA accepting *L*, i.e.,  $\mathcal{L}_m(G) = L$ . We also make the following assumptions:

A-1  $\mathcal{L}(G)$  is live; and

A-2 G does not contain an unobservable cycle; and

A-3  $\mathcal{L}(G) = \overline{\mathcal{L}_m(G)}$ , i.e., *G* is non-blocking.

Assumptions A-1 and A-2 are standard assumptions in the analysis of partially-observed DES in order to avoid trivial result. We make assumption A-3 just to simplify the technical development and it is without of loss generality, since we can take the trim part of *G*, i.e., removing all states that cannot reach a mark state from *G*, if  $\mathcal{L}(G) \neq \overline{\mathcal{L}_m(G)}$ .

#### 4.1. Twin-Machine and information merge

In order to verify trajectory detectability, we construct a new verification DFA  $V = (X_V, \Sigma_V, f_V, x_{0,V}, X_{m,V})$ , where

- $X_V \subseteq Q \times Q$  is the set of states;
- $\Sigma_v = (\Sigma_o \times \Sigma_o) \cup (\Sigma_{uo} \times \{\epsilon\}) \cup (\{\epsilon\} \times \Sigma_{uo})$  is the set of events;
- $x_{0,V} = (q_0, q_0)$  is the initial state;
- $X_{m,V} = Q_m \times Q_m$  is the set of marked states;
- $f_V : X_V \times \Sigma_v \to X_V$  is the partial transition function defined by: for any state  $(x_1, x_2) \in X_V$  and an event  $\sigma \in \Sigma$ , we have
  - (a) If  $\sigma \in \Sigma_0$ , then the following transition is defined:

$$f_V((x_1, x_2), (\sigma, \sigma)) = (\delta(x_1, \sigma), \delta(x_2, \sigma))$$
(9)

(b) If  $\sigma \in \Sigma_{uo}$ , then the following transitions are defined:

 $f_V((x_1, x_2), (\sigma, \epsilon)) = (\delta(x_1, \sigma), x_2)$ (10)

$$f_V((x_1, x_2), (\epsilon, \sigma)) = (x_1, \delta(x_2, \sigma))$$
(11)

Hereafter, we only consider the reachable part of *V*.

**Remark 5.** The construction of DFA *V* essentially follows the idea of the well-known twin-machine (or verifier) construction used in the literature for the verification of diagnosability; see, e.g., [21,22]. Roughly speaking, *V* tracks all pairs of observational equivalent strings in *G*. Specifically, if  $s_1, s_2 \in \mathcal{L}(G)$  are two strings in *G* such that  $P(s_1) = P(s_2)$ , then there exists a string  $s \in \mathcal{L}(V)$  in *V* such that its first and second components are  $s_1$  are  $s_2$ , respectively, i.e.,  $f_V(x_{0,V}, s) = (\delta(q_0, s_1), \delta(q_0, s_2))$ . On the other hand, for any string  $s = (s_1, s_2) \in \mathcal{L}(V)$  in *V*, we have that  $P(s_1) = P(s_2)$ .

Before we show how to use DFA *V* to verify trajectory detectability, let us consider the following scenario. Suppose that there exist two *different* strings  $s_1, s_2 \in \mathcal{L}(G)$  such that  $P(s_1) = P(s_2)$  and  $\delta(q_0, s_1) = \delta(q_0, s_2)$ . Since these two strings lead to the same state, we know that they have the same continuations. Moreover, since  $s_1$  and  $s_2$  are observational equivalent, we know that we cannot distinguish any continuations of these two strings from this state. We call such a phenomenon *information merge*. Clearly, if the information merge phenomenon occurs, then  $\mathcal{L}_m(G)$  becomes non-detectable. Therefore, we need to check whether or not the information merge phenomenon occurs as the first step towards the verification of trajectory detectability. To the end, we first introduce the notion of merging state.

**Definition 5.** A state  $x = (x_1, x_2) \in X_V$  is said to be a *merging state* if

$$[x_1 \neq x_2] \land [\exists \sigma \in \Sigma_0 : \delta(x_1, \sigma) = \delta(x_2, \sigma)]$$
(12)

or

$$\begin{bmatrix} In^{V}(x) \cap (\Sigma_{o} \times \Sigma_{o}) \neq \emptyset \lor x = x_{0,V} \end{bmatrix}$$
(13)  
 
$$\wedge \left[ \exists e_{1}, e_{2} \in \Sigma_{uo}^{*} : [\delta(x_{1}, e_{1}) = \delta(x_{2}, e_{1})] \land [x_{1} = x_{2} \Rightarrow e_{1} \neq e_{2}] \right]$$

We denote by  $X_{merge} \subseteq X_V$  the set of merging states.

Intuitively, a merging state  $(x_1, x_2)$  is either the initial state  $x_{0,V}$  or a state reached immediately after a pair of observable events such that  $x_1$  and  $x_2$  can reach a same state unobservably or via a single common observable event. Moreover, the strings from  $x_1$  and  $x_2$  to the same state should be different if  $x_1 \neq x_2$ . The following result reveals that the information merge phenomenon will not happen in *G* if and only if *V* does not contain a merging state.

**Theorem 1.** There exist two different strings  $s_1, s_2 \in \mathcal{L}(G)$  such that  $P(s_1) = P(s_2)$  and  $\delta(q_0, s_1) = \delta(q_0, s_2)$  if and only if  $X_{merge} \neq \emptyset$ .

**Proof.** ( $\Leftarrow$ ) Suppose that  $X_{merge} \neq \emptyset$ . This means that there exists a state  $(x_1, x_2) \in X_V$  such that either Eq. (12) holds or Eq. (13) holds. Let  $s = (s_1, s_2)$  be a string such that  $f_V(x_{0,V}, s) = (x_1, x_2)$ , where we know that  $P(s_1) = P(s_2)$  and  $s_1, s_2 \in \mathcal{L}(G)$ . We define two new strings  $s'_1$  and  $s'_2$  such that

$$s'_{i} = \begin{cases} s_{i}\sigma_{0} & \text{if Eq. (12) holds} \\ s_{i}e_{i} & \text{if Eq. (13) holds} \end{cases}$$
(14)

where  $i = 1,2, \sigma_0$  is an event satisfying Eq. (12) and  $e_1, e_2 \in \Sigma_{uo}^*$ are strings satisfying Eq. (13). Clearly, if Eq. (12) holds, we know that  $P(s'_1) = P(s_1)\sigma_0 = P(s_2)\sigma_0 = P(s'_2)$ . Similarly, if Eq. (13) holds, we also have that  $P(s'_1) = P(s_1) = P(s_2) = P(s'_2)$ . Moreover, since either  $x_1 \neq x_2$  or  $e_1 \neq e_2$ , we know that  $s'_1 \neq s'_2$ , i.e., they are two different strings. Overall, we know that  $P(s'_1) = P(s'_2)$  and  $\delta(q_0, s'_1) = \delta(q_0, s'_2)$ .

(⇒) Suppose that there exist two different strings  $s_1, s_2 \in \mathcal{L}(G)$  such that  $P(s_1) = P(s_2)$  and  $\delta(q_0, s_1) = \delta(q_0, s_2)$ . We denote by  $l_1 \leq s_1$  the shortest prefix of  $s_1$  such that

$$\exists l' \leq s_2 : P(l_1) = P(l') \land l_1 \neq l' \land \delta(q_0, l_1) = \delta(q_0, l')$$

$$(15)$$

Note that  $l_1$  is well-defined, since Eq. (15) always holds by taking  $l_1 = s_1$  and  $l' = s_2$ . We denote by  $l_2$  a string satisfying Eq. (15) for  $l_1$ . Also, we know that  $l_1 \neq \epsilon$ ; otherwise, it implies that there exists an unobservable cycle at  $q_0$  in *G*, which contradicts the assumption. Similarly, we know that  $l_2 \neq \epsilon$ . Therefore, we write  $l_1 = \hat{l}_1 \sigma_1$  and  $l_2 = \hat{l}_2 \sigma_2$ , where  $\sigma_1$  and  $\sigma_2$  are the last events in  $l_1$  and  $l_2$ , respectively. Then we consider the following four cases for  $\sigma_1$  and  $\sigma_2$ :

Case 1:  $\sigma_1, \sigma_2 \in \Sigma_0$ .

In this case, first, we know that  $\sigma_1 = \sigma_2$ ; otherwise,  $P(l_1) \neq P(l_2)$ . Second, we know that  $\delta(q_0, \hat{l}_1) \neq \delta(q_0, \hat{l}_2)$ ; otherwise  $\hat{l}_1$  will be the shortest prefix of  $s_1$  satisfying Eq. (15). Therefore, since  $P(\hat{l}_1) = P(\hat{l}_2)$ , we know that state  $x := (\delta(q_0, \hat{l}_1), \delta(q_0, \hat{l}_2))$  is reachable in *V*. Hence, state *x* and event  $\sigma := \sigma_1 = \sigma_2$  satisfy Eq. (12), i.e.,  $x \in X_{merge} \neq \emptyset$ .

Case 2:  $\sigma_1 \in \Sigma_o$  and  $\sigma_2 \in \Sigma_{uo}$ .

In this case, we know that the last observable event in  $\hat{l}_2$  must be  $\sigma_1$ ; otherwise,  $P(l_1) \neq P(l_2)$ . Therefore, we can write  $l_2$  by  $l_2 = w_2\sigma_1\xi_2$ , where  $\xi_2 \in \Sigma_{uo}^*$ . Also, we know that  $\delta(q_0, \hat{l}_1\sigma_1) \neq \delta(q_0, w_2\sigma_1)$ ; otherwise,  $\xi_2$  yields an unobservable cycle at  $\delta(q_0, w_2\sigma_1)$ . Since  $P(\hat{l}_1\sigma_1) = P(w_2\sigma_1)$ , we know that  $x := (\delta(q_0, \hat{l}_1\sigma_1), \delta(q_0, w_2\sigma_1))$  is reachable in *V*. Moreover,  $(\sigma_1, \sigma_1)$   $\in$  In<sub>V</sub>(*x*), i.e., In<sub>V</sub>(*x*) $\cap$ ( $\Sigma_0 \times \Sigma_0$ )  $\neq \emptyset$ . Hence, state *x* satisfies Eq. (13) by taking  $e_1 = \epsilon$  and  $e_2 = \xi_2$  and we know that  $x \in X_{merge} \neq \emptyset$ .

Case 3:  $\sigma_1 \in \Sigma_{uo}$  and  $\sigma_2 \in \Sigma_o$ .

This case is analogous to Case 2 and we still can show that  $X_{merge} = \emptyset$ .

Case 4:  $\sigma_1 \in \Sigma_{uo}$  and  $\sigma_2 \in \Sigma_{uo}$ .

First, suppose that  $P(l_1) = P(l_2) = \epsilon$ . Then we can take  $(q_0, q_0) = x_{0,V}$  and we know that  $\delta(q_0, l_1) = \delta(q_0, l_2)$  and  $l_1 \neq l_2$ , i.e., Eq. (13) is satisfied. Next, suppose that  $P(l_1) = P(l_2) \neq \epsilon$ . Let  $\sigma$  be the last observable event in  $l_1$  and  $l_2$  and we write  $l_1 = w_1\sigma\xi_1$  and  $l_2 = w_2\sigma\xi_2$ , where  $\xi_1, \xi_2 \in \Sigma_{uo}^*$ . First, we know that  $\delta(q_0, w_1\sigma) \neq \delta(q_0, w_2\sigma)$ ; otherwise  $w_1\sigma$  will be the shortest prefix of  $s_1$  satisfying Eq. (15). Since  $P(w_1\sigma) = P(w_2\sigma)$ , we know that state  $x = (\delta(q_0, w_1\sigma), \delta(q_0, w_2\sigma))$  is reachable in *V*. Moreover,  $(\sigma, \sigma) \in \ln_V(x)$ , i.e.,  $\ln_V(x) \cap (\Sigma_o \times \Sigma_o) \neq \emptyset$ . Therefore, by taking  $x = (\delta(q_0, w_1\sigma), \delta(q_0, w_2\sigma))$ ,  $e_1 = \xi_1$  and  $e_2 = \xi_2$ , we know that Eq. (13) is satisfied, i.e.,  $x \in X_{merge} \neq \emptyset$ .  $\Box$ 

#### 4.2. Verification of trajectory detectability

Let  $x \in X_V$  be a state in *V*. We say that *x* is in a cycle of *V* if there exists a non-empty string  $s \in \Sigma_V^*$  such that  $f_V(x, s) = x$ . We denote by  $C_V \subseteq X_V$  the set of states in some cycle of *V*.

Now, we are ready to show how to verify trajectory detectability by using DFA V.

**Theorem 2.** Let  $G = (Q, \Sigma, \delta, q_0, Q_m)$  be a DFA and V be the DFA constructed from G. Then  $\mathcal{L}_m(G)$  is trajectory detectable w.r.t.  $\Sigma_o$  if and only if  $X_{merge} = \emptyset$  and

$$Acc_{V}(\mathcal{C}_{V}) \cap (X_{m,V} \setminus \{(q,q) : q \in Q_{m}\}) = \emptyset$$
(16)

That is, V does not contain a merging state and any cycle in V cannot reach a marked state in which the first and the second components are not identical.

**Proof.**  $(\Rightarrow)$  By contraposition. First, suppose that Eq. (16) does not hold. Let  $x = (x_1, x_2)$  be a state in  $Acc_V(\mathcal{C}_V) \cap (X_{m,V} \setminus \{(q, q) : q \in Q_m\})$ . Since  $x \in Acc_V(\mathcal{C}_V)$ , we know that there exists a string

$$\underbrace{\sigma_0\sigma_1\ldots\sigma_p}_{=:s}\underbrace{\sigma_{p+1}\ldots\sigma_{p+m}}_{=:w}\underbrace{\sigma_{p+m+1}\ldots\sigma_{p+m+k}}_{=:t}\in\mathcal{L}(V)$$

such that

1. 
$$f_V(x_{0,V}, sw^n t) = x$$
 for any  $n \in \mathbb{N}$ ; and  
2.  $\exists i = 1, ..., m : \sigma_{n+i} \in \Sigma \times \Sigma$ .

The first condition says that string w forms a cycle and the second condition says that w must contain an event in the form of  $(\sigma, \sigma), \sigma \in \Sigma_0$ . This comes from Assumption A-2, since we know that a state in  $C_V$  can never reach itself by a string in  $((\Sigma_{uo} \times \{\epsilon\}) \cup (\{\epsilon\} \times \Sigma_{uo}))^*$ ; otherwise, it implies that there exists an unobservable cycle in G. For strings s, w and t, we denote by  $s_i$ ,  $w_i$  and  $t_i, i = 1, 2$  their ith components, respectively, i.e., swt = $(s_1, s_2)(w_1, w_2)(t_1, t_2)$ . We know that  $P(s_1) = P(s_2), P(w_1) = P(w_2)$ and  $P(t_1) = P(t_2)$ . Moreover, we know that  $|P(w_1)| = |P(w_2)| \ge 1$ . Since  $x \in X_m \setminus \{(q, q) : q \in Q_m\}$ , we know that  $s_i w_i^n t_i \in \mathcal{L}_m(G), \forall i = 1, 2, \forall n \in \mathbb{N}$  and  $s_1 w_1^n t_1 \neq s_2 w_2^n t_2, \forall n \in \mathbb{N}$ . Therefore, for any  $n \in \mathbb{N}$ , we can find a string  $s_1 w_1^n t_1 \in \mathcal{L}_m(G)$  such that  $|P(s_1 w_1^n t_1)| \ge |P(w_1^n)| \ge 1$  and  $P^{-1}P(s_1 w_1^n t_1) \supseteq \{s_1 w_1^n t_1, s_2 w_2^n t_2\}$ , i.e., Card $[P^{-1}P(s_1 w_1^n t_1) \cap \mathcal{L}_m(G)] > 1$ . This implies that  $\mathcal{L}_m(G)$  is not trajectory detectable.

Next, we suppose that  $X_{merge} \neq \emptyset$ . By Theorem 1, we know that there exist two different strings  $s_1, s_2 \in \mathcal{L}(G)$  such that  $P(s_1) = P(s_2)$  and  $\delta(q_0, s_1) = \delta(q_0, s_2) =: q$ . Therefore, for any  $n \in \mathbb{N}$ , we can choose a string w defined at q, such that  $\delta(q, w) \in X_m$ ,  $|w| \ge n$  and  $|P(w)| \ge n$ . Such a string w is well-defined since we assume



(a) Part of the DFA V for  $G_1$  in Figure 1(a).



(b) The DFA V for  $G_4$  in Figure 2(a).

Fig. 3. Examples of the verification DFA V.

that *G* is live and non-blocking and there is no unobservable cycle in *G*. For the above  $s_1, s_2$  and w, we know that  $s_1w, s_2w \in \mathcal{L}_m(G)$ and  $P(s_1w) = P(s_2w)$ , i.e.,  $Card[P^{-1}P(s_1w) \cap \mathcal{L}_m(G)] > 1$ . This implies that  $\mathcal{L}_m(G)$  is not trajectory detectable.

(⇐) Still by contraposition. Suppose that  $\mathcal{L}_m(G)$  is not trajectory detectable. This implies that there exist two different arbitrarily long strings in  $\mathcal{L}_m(G)$  having the same projection. Since the state-space of *G* is finite, any arbitrarily long string must be contributed by some cycle. Therefore, we can find strings  $t_i$ ,  $w_i$ ,  $v_i \in \Sigma^*$ , i = 1, 2 such that  $P(t_1) = P(t_2)$ ,  $P(w_1) = P(w_2)$ ,  $P(v_1) = P(v_2)$  and  $\delta(q_0, t_1w_1^nv_1)$ ,  $\delta(q_0, t_2w_2^nv_2) \in X_m$  for any  $n \in \mathbb{N}$ , i.e.,  $t_1w_1^nv_1$ ,  $t_2w_2^nv_2 \in \mathcal{L}_m(G)$ . Next, we consider two cases:

Case 1:  $\delta(q_0, t_1w_1v_1) \neq \delta(q_0, t_2w_2v_2)$ ; Case 2:  $\delta(q_0, t_1w_1v_1) = \delta(q_0, t_2w_2v_2)$ .

First, assume that Case 1 holds. Let us consider the state reached by  $sw := (s_1, s_2)(w_1, w_2)$  in V, say  $x_w$ . Since  $sw^n \in \mathcal{L}(V)$  for any  $n \in \mathbb{N}$ , we know that  $x_w \in \mathcal{C}_V$ . Since  $x_v := (\delta(q_0, t_1w_1v_1), \delta(q_0, t_2w_2v_2)) \in X_{m,V}$  and  $\delta(q_0, t_1w_1v_1) \neq \delta(q_0, t_2w_2v_2)$ , we know that  $x_v \in X_m \setminus \{(q, q) : q \in Q_m\}$ . Moreover, since state  $x_w$  can reach state  $x_v$ via string  $v = (v_1, v_2)$ , we know that  $x_v \in Acc_V(\mathcal{C}_V) \cap (X_m \setminus \{(q, q) : q \in Q_m\}) \neq \emptyset$ . Second, we assume that Case 2 holds. Then, since  $t_1w_1v_1 \neq t_2w_2v_2$ ,  $P(t_1w_1v_1) = P(t_2w_2v_2)$  and  $\delta(q_0, t_1w_1v_1) = \delta(q_0, t_2w_2v_2)$ , by Theorem 1, we know immediately that  $X_{merge} \neq \emptyset$ . This completes the contraposition proof.  $\Box$ 

Let us illustrate how to use Theorem 2 to verify trajectory detectability by the following example.

**Example 4.** First, let us consider language  $\mathcal{L}(G_1)$  generated by DFA  $G_1$  shown in Fig. 1(a). Note that, since the definition of trajectory detectability considers marked languages, we need to mark all states in  $G_1$  such that  $\mathcal{L}_m(G_1) = \mathcal{L}(G_1)$ . Therefore, all states in V are also marked states. Then part of the verification DFA V for  $G_1$  is shown in Fig. 3(a). Note that the self-loop at state  $(5, 5) \in X_{m,V}$  does not violate Eq. (16), since the first and the second components in (5, 5) are the same. However, state (3, 4) is a merging state, since (i) it is entered by  $(o, o) \in \Sigma_o \times \Sigma_o$ , i.e.,  $\ln_V((3, 4)) \cap (\Sigma_o \times \Sigma_o) \neq \emptyset$ ; and (ii) state 3 and 4 are different states; and (iii) event  $(o, o) \in$  $\Sigma_0 \times \Sigma_0$  is defined at (3, 4) and it leads to state (5, 5) whose two components are the same state. Therefore, we know that  $\mathcal{L}(G_1)$  is not trajectory detectable. Note that, since trajectory detectability does not depend on the DFA generating the language, one can also construct the verification DFA V based on  $G_2$  shown in Fig. 1(b), which will give the same result.

Next, let us consider language  $\mathcal{L}(G_4)$  marked by DFA  $G_4$  shown in Fig. 2(a). The complete verification DFA V for  $G_4$  is shown in Fig. 3(b). First, we can easily check that there is no merging state in V, i.e.,  $X_{merge} = \emptyset$ . Second, for states that can be reached from

a cycle in *V*, i.e., states (1, 1), (3, 3), (1, 2), (3, 4), (2, 2) and (4, 4), they are either unmarked states or marked states but in the form of (q, q). Therefore, we know that  $\mathcal{L}_m(G_4)$  is trajectory detectable. However, if we want to consider the language generated by  $G_4$ , i.e.,  $\mathcal{L}(G_4)$ , then we need to mark all states in  $G_4$ . In this case, states (1, 2) and (3, 4) become marked states and they are still in a cycle. Therefore, we know that the generated language  $\mathcal{L}(G_4)$  is not trajectory detectable.  $\Box$ 

**Remark 6.** Let us explain the intuition behind Theorem 2. First, the condition in Eq. (16) requires that, for any marked state that can be reached from a cycle in *V*, it must be in the form of  $\{(q, q) : q \in Q_m\}$ . Otherwise, suppose that a state  $(q_1, q_2) \in Q_m \times Q_m, q_1 \neq q_2$  is reached from a cycle. Then we know that there exist two arbitrarily long strings in  $\mathcal{L}_m(G)$  leading to  $q_1$  and  $q_2$ , respectively, such that they have the same observation. Moreover,  $s_1 \neq s_2$  since  $q_1 \neq q_2$ . However, only requiring this condition is not sufficient to recover the trajectory, since it is possible that all states in some cycle are in the form of  $\{(q, q) : q \in Q_m\}$  but two strings may merge to a same state before they enter some cycle. If such a scenario occurs, then we can identify the precise trajectory leading to this state. This is why we need that  $X_{merge} \neq \emptyset$  in addition to Eq. (16).

**Remark 7.** We discuss the complexity for verifying trajectory detectability by using Theorem 2. Let  $G = (Q, \Sigma, \delta, q_0, Q_m)$  be the DFA for which we want to verify whether or not  $\mathcal{L}_m(G)$  is trajectory detectable. First, we need to construct the verification DFA *V*, which takes  $O(|\Sigma||Q|^2)$ , since there are at most  $|Q|^2$  states and  $3|\Sigma||Q|^2$  transitions in *V*. Then for each state in *V*, it takes  $|\Sigma||Q|$  to determine whether it is a merging state or not. Therefore, it takes  $O(|\Sigma|^2|Q|^3)$  to check whether or not  $X_{merge} = \emptyset$ . Finally, we need to check whether or not there exists a state violating Eq. (16). To this end, we can first compute  $C_V$ , which can be done in  $O(|\Sigma||Q|^2)$  by using Kosaraju's algorithm; see, e.g., [27]. Computing the reachable set  $Acc_V(C_V)$  still requires  $O(|\Sigma||Q|^2)$ . Overall, the total complexity of the verification of trajectory detectability is  $O(|\Sigma|^2|Q|^3)$ , which is polynomial w.r.t. the size of G.

#### 4.3. Verification of periodic trajectory detectability

Now, let us investigate the verification of periodic trajectory detectability. In contrast to the verification of trajectory detectability, we cannot use the verification DFA V to verify periodic trajectory detectability. Instead, we use the observer structure [24].

Let  $G = (Q, \Sigma, \delta, q_0, Q_m)$  be a DFA. The observer of G w.r.t.  $\Sigma_o$ is a DFA  $Obs(G) = (X_{obs}, \Sigma_o, f_{obs}, x_{0,obs})$ , where  $X_{obs} \subseteq 2^Q$ ,  $x_{0,obs} = R_G(\epsilon)$  and  $f_{obs} : X_{obs} \times \Sigma_o \to X_{obs}$  such that, by extending it to  $X_{obs} \times \Sigma_o^*$ , we have  $f_{obs}(x_{0,obs}, s) = R_G(s)$  for any  $s \in P(\mathcal{L}(G))$ .

Let  $x \in X_{obs}$  be a state in the observer. We say that x is

- a certain state if  $\exists q \in x : q \in Q_m \land x \setminus \{q\} \subseteq Q \setminus Q_m$ ;
- an uncertain state if  $\exists q_1, q_2 \in x : q_1, q_2 \in Q_m \land q_1 \neq q_2$ .

We denote by  $X_{obs}^{ct}$  and  $X_{obs}^{uc}$  the set of certain states and the set of uncertain states, respectively. Intuitively, a certain state only contains one marked state and a uncertain state contains some different marked states. Also, we say that a set of distinct states  $\{x_1, \ldots, x_n\} \subseteq X_{obs}$  is a cycle in Obs(G) if we can find a sequence of events  $\sigma_1, \ldots, \sigma_n$  such that  $f_{obs}(x_i, \sigma_i) = x_{i+1}, \forall i = 1, \ldots, n$ , where  $x_{n+1} := x_1$ .

The following result reveals how to use the observer to verify periodic trajectory detectability.

**Theorem 3.** Let  $G = (Q, \Sigma, \delta, q_0, Q_m)$  be a DFA and Obs(G) be the observer w.r.t.  $\Sigma_o$ . Then  $\mathcal{L}(G)$  is periodically trajectory detectable w.r.t.  $\Sigma_o$  if and only if

- (i)  $X_{merge} = \emptyset$ ; and
- (ii) for any cycle  $C \subseteq X_{obs}$  in Obs(G), if  $C \cap X_{obs}^{ct} = \emptyset$ , then any state in C cannot reach a state in  $X_{obs}^{uc}$  through states in  $X_{obs} \setminus X_{obs}^{ct}$ .

**Proof.**  $(\Rightarrow)$  By contraposition. Suppose that  $X_{merge} \neq \emptyset$ . We know that there exist two strings  $s_1, s_2 \in \mathcal{L}(G)$  such that  $P(s_1) = P(s_2)$ and  $\delta(q_0, s_1) = \delta(q_0, s_2) =: q$ . Let  $\hat{s}_1 \leq s_1$  be the longest prefix of  $s_1$  such that  $\hat{s}_1 \in \mathcal{L}_m(G)$  and we write  $s_1 = \hat{s}_1 s'_1$ . Let w be an arbitrarily long string from q to a marked state. Then for any  $n \in \mathbb{N}$ , there exists a string  $\hat{s}_1 \in \mathcal{L}_m(G)$  and a continuation  $s'_1 w \in \mathcal{L}_m(G) / \hat{s}_1$ such that  $|P(s'_1 w)| \geq n$ . Moreover, for any  $w' \leq w : \hat{s}s'_1 w \in \mathcal{L}_m(G)$ , we have  $\operatorname{Card}[P^{-1}P(\hat{s}_1 s'_1 w') \cap \mathcal{L}_m] > 1$ , since  $s_2 w' \neq \hat{s}_1 s'_1 w'$  and  $P(\hat{s}_1 s'_1 w') = P(s_1 w') = P(s_2 w')$ . This implies that  $\mathcal{L}_m(G)$  is not periodically trajectory detectable.

Now, let us suppose that there exists a cycle *C* such that  $C \cap X_{obs}^{ct} = \emptyset$  and a state in it can reach a state in  $X_{obs}^{uc}$  through states in  $X_{obs} \setminus X_{obs}^{ct}$ . Let  $x \in C$  be such a state in the cycle. Let  $s_o \in \Sigma_o^*$  be a string that reaches x from  $x_{0,obs}$ . Let  $w_o \in \Sigma_o^*$  be a string such that  $f_{obs}(x, w_o) = x$  and it only visits all states in *C*. Let  $t_o \in \Sigma_o^*$  be a string that reaches a state in  $X_{obs}^{uc}$  from x through states in  $X_{obs} \setminus X_{obs}^{ct}$ . Formally, we have that  $f_{obs}(x, t_o) \in X_{obs}^{uc}$  and  $\forall t'_o \leq t_o : f_{obs}(x, t'_o) \notin X_{obs}^{ct}$ . By the property of the observer, we know that, for any  $n \in \mathbb{N}$ , there exists a string  $sw^n t \in P^{-1}(s_ow_o^n t_o)$  such that  $P(s) = s_o$ ,  $P(w) = w_o$ ,  $P(t) = t_o$  and  $\delta(q_0, sw^n t) \in Q_m$ . Still, we denote by  $\hat{s} \leq s$  the longest prefix of s such that  $\hat{s} \in \mathcal{L}_m(G)$  and a continuation  $s'w^n t \in \mathcal{L}_m(G)/\hat{s}$  such that  $|P(s'w^n t)| \geq n$ . Moreover, for any  $w' \leq w^n t : \hat{ss'}w' \in \mathcal{L}_m(G)$ , we have Card $[P^{-1}P(\hat{ss'}w') \cap \mathcal{L}_m(G)] > 1$ , since  $f_{obs}(x_{0,obs}, P(\hat{s}_1s'w'))$  is always not a certain state.

(⇐) Still by contraposition. Suppose that  $\mathcal{L}_m(G)$  is not periodically trajectory detectable. This implies that there exists a string  $s \in \mathcal{L}_m(G)$  and an arbitrarily long continuation  $u \in \mathcal{L}_m(G)/s$  such that  $\forall u' \leq u : su' \in \mathcal{L}_m(G)$ , we have that  $\operatorname{Card}[P^{-1}P(su') \cap \mathcal{L}_m(G)] > 1$ . Since the state-spaces of *G* and *Obs*(*G*) are both finite, any arbitrarily long string and its projection must be contributed by some cycle in *G* and some cycle in *Obs*(*G*), respectively. Therefore, we can find string *swt* such that  $\forall n \in \mathbb{N}, \delta(q_0, sw^n t) \in \mathcal{L}_m(G), f_{obs}(x_{0,obs}, P(s)P(w)^n P(t))!$  and  $(\forall w' \leq w^n t : sw' \in \mathcal{L}_m(G))[\operatorname{Card}[P^{-1}P(sw') \cap \mathcal{L}_m(G)] > 1].$ 

Next, we consider the following two cases. First, assume that

$$(\exists w' \le wt : sw' \in \mathcal{L}_m(G))(\exists s_2 \in \mathcal{L}_m(G))$$

$$[s_2 \ne sw' \land P(sw') = P(s_2) \land \delta(q_0, sw') = \delta(q_0, s_2)]$$
(17)

Then, by Theorem 1, we know that  $X_{merge} \neq \emptyset$ . Therefore, hereafter, we assume that

$$(\forall w' \le wt : sw' \in \mathcal{L}_m(G))(\forall s_2 \in \mathcal{L}_m(G)) \text{ s.t.}$$
(18)  
$$[s_2 \ne sw' \land P(sw') = P(s_2)] \Rightarrow [\delta(q_0, sw') \ne \delta(q_0, s_2)]$$

Let  $x_{sw} := f_{obs}(x_{0,obs}, P(sw))$  and  $x_{swt} := f_{obs}(x_{0,obs}, P(swt))$ . First, by Card $[P^{-1}P(swt) \cap \mathcal{L}_m(G)] > 1$  and Eq. (18), we know that  $x_{swt} \in X_{obs}^{uc}$ . Moreover, for any  $w' \le wt$ , we know that  $f_{obs}(x_{0,obs}, P(sw')) \notin X_{obs}^{ct}$ , since if it contains one state in  $Q_m$ , then it must contain a distinct state in  $Q_m$ . Therefore, we know that  $x_{sw}$  is in a cycle that does not contain a certain state and  $x_{sw}$  can reach uncertain state  $x_{swt}$  without reaching certain states. This completes the contrapositive proof.  $\Box$ 

**Example 5.** Let us consider  $\mathcal{L}_m(G_5)$  marked by DFA  $G_5$  shown in Fig. 2(b) with  $\Sigma_o = \{o\}$ . First, we can show that  $X_{merge} = \emptyset$ . Then the observer  $Obs(G_5)$  is shown in Fig. 4(a). We know that state  $\{4, 5, 6\}$  is a certain state and state  $\{1, 2, 3\}$  is an uncertain state. However, these two states form the only cycle in  $Obs(G_5)$  and one of which is a certain state. Therefore, we know that  $\mathcal{L}_m(G_5)$  is periodically trajectory detectable. Note that, if we consider  $\mathcal{L}(G_5)$ ,



Fig. 4. Examples of the verification DFA V.

i.e, we need to mark all states in  $G_5$ , then both  $\{1, 2, 3\}$  and  $\{4, 5, 6\}$  are uncertain states, which means that the generated language  $\mathcal{L}(G_5)$  is not periodically trajectory detectable.  $\Box$ 

**Remark 8.** Note that, in the worst case, the size of the observer is exponential in the size of *G*. To implement Theorem 3, first, we need to check whether or not  $X_{merge} = \emptyset$ . This can still be done by using the verification DFA *V* and it takes  $O(|\Sigma|^2|Q|^3)$ . Determining whether or not there exists a cycle that does not contain a certain state can be done in  $O(|X_{obs}||\Sigma_o|)$ . Checking whether or not there exists a state in such a cycle that can reach an uncertain state without visiting certain states can still be done in  $O(|X_{obs}||\Sigma_o|)$ by a depth-first search. Therefore, the complexity for verifying periodic trajectory detectability is  $O(|\Sigma|2^{|Q|} + |\Sigma|^2|Q|^3)$ , which is exponential in the size of *G*.

**Remark 9.** Let us explain why DFA *V* cannot be used to check periodic trajectory detectability. Let us consider again the system  $G_5$  studied in Example 5 and we construct part of the verification DFA *V* shown in Fig. 4(b). However, we cannot determine the "certainty" of a trajectory based on *V*. For example, state (4, 5), where 5 is a marked state, corresponds to certain state {4, 5, 6} in  $Obs(G_5)$ . However, state (1, 2), which also only contains a single marked state, corresponds to uncertain state {1, 2, 3} in  $Obs(G_5)$ . This is because, although state (1, 2) looks like "certain" in the leftmost cycle, the marked state in it can be confused with other marked state in a different cycle, e.g., the rightmost cycle in Fig. 4(b). Therefore, exponential complexity arises when this combination is considered. One interesting future direction is to identify the precise complexity class for the verification of periodic trajectory detectability.

#### 5. Conclusion

We investigated the trajectory estimation problem for partiallyobserved DES. A language-based framework was proposed to study this problem. The notions of trajectory detectability and periodic trajectory detectability were proposed in order to capture different requirements in practice. We provided effective algorithms to verify these notions when the system language is regular.

There are several future directions for the proposed framework. First, in this paper, we only consider to verify a priori if the trajectory of a system can be detected. Investigating effective online detection mechanism is also an important future direction. Second, it is important to investigate the decidability of trajectory detectability for other classes of languages, e.g. Petri nets languages [28]. Finally, the observation mapping considered in this paper is static. Extending the proposed framework to the dynamic observation setting [29,30] or the non-deterministic observation setting [31–33] is also an interesting direction.

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