On Approximate Opacity of Cyber-Physical Systems

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Abstract—Opacity is an important information-flow security property in the analysis of cyber-physical systems. It captures the plausible deniability of the system’s secret behavior in the presence of an intruder that may access the information flow. Existing works on opacity only consider non-metric systems by assuming that the intruder can always distinguish two different outputs precisely. In this paper, we extend the concept of opacity to systems whose output sets are equipped with metrics. Such systems are widely used in the modeling of many real-world systems whose measurements are physical signals. A new concept called approximate opacity is proposed in order to quantitatively evaluate how close two outputs are. Such systems allows us to verify approximate opacity for large-scale, or even infinite systems, using their abstractions. We also discuss how to construct approximate opacity preserving symbolic models for a class of discrete-time control systems. Our results extend the definitions and analysis techniques for opacity from non-metric systems to metric systems.

Index Terms—Opacity; Approximate Simulation Relations; Finite Abstractions; Symbolic Models.

I. INTRODUCTION

A. Motivations

Cyber-physical systems (CPS) are complex systems resulting from tight interactions of dynamical systems and computational devices. Such systems are generally very complex posing both continuous and discrete behaviors which makes the verification and design of such systems significantly challenging. In particular, components in CPS are usually connected via communication networks in order to acquire and exchange information so that some global functionality of the system can be achieved. However, this also brings new challenges for the verification and design of CPS since the communication between system components may release information that might compromise the security of the system.

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Therefore, how to analyze and enforce security for CPS is becoming an increasingly important issue and has drawn considerable attention in the literature in the past few years [1], [2].

In this paper, we investigate an important information-flow security property called opacity. Roughly speaking, opacity is a confidentiality property that captures whether or not the “secret” of the system can be revealed to an intruder that can infer the system’s actual behavior based on the information flow. A system is said to be opaque if it always has the plausible deniability for any of its secret behavior. The concept of opacity was originally proposed in the computer science literature as a unified notion for several security properties [3], [4]. Since then, opacity has been studied more extensively in the context of Discrete-Event Systems (DES), an important class of event-driven dynamical systems with discrete state spaces. For example, in [5]–[7], several state-based notions of opacity were proposed, which include current-state opacity, initial-state opacity, K-step opacity and infinite-step opacity. In [8], the author proposed two language-based opacity notions called strong opacity and weak opacity and investigated their relationships with some other properties. In [9], transformation algorithms among different notions of opacity were proposed. The above mentioned works mainly consider DES modeled by finite-state automata. More recently, the definitions and verification algorithms for different notions of opacity have been extended to other classes of (discrete) systems, including Petri nets [10]–[13], stochastic systems [14]–[16], recursive tile systems [17] and pushdown systems [18]. The interested readers are referred to recent surveys [19], [20] for more references and recent developments on this active research area.

Since opacity is an information-flow property, its definition strictly depends on the information model of the system. Most of the existing works in the literature formulate opacity by adopting the event-based observation model, i.e., some events of the system (either transition labels or state labels) are observable or distinguishable while some are not. This essentially assumes that the output of the system is symbolic in the sense that we can precisely distinguish two outputs with different labels. Hereafter, we will also refer to opacity under this setting as exact opacity. Exact opacity is very meaningful for systems whose output sets are non-metric, e.g., discrete systems whose outputs are logic events. However, for many real-world applications whose outputs are physical signals, instead of just saying that two events are distinguishable or indistinguishable, we may have a measurement to quantitatively evaluate how close two outputs are. Such systems
are referred to as metric systems, where the output sets are equipped with appropriate metrics. For metric systems, if two signals are very close to each other, then it will be very hard to distinguish them unambiguously due to the measurement precision or potential measurement noises. A typical example of this scenario is linear or nonlinear discrete-time control systems with continuous state-spaces and continuous output mappings. Therefore, existing definitions of opacity are too strong for metric systems since they implicitly assume that the intruder can always distinguish two output signals even when they are arbitrarily close to each other, which is not practical.

### B. Our Contributions

In this paper, we propose a new concept called approximate opacity that is more applicable to metric systems. In particular, we treat two outputs as “indistinguishable” outputs if their distance is smaller than a given threshold parameter $\delta \geq 0$. We consider three basic types of opacity, initial-state opacity, current-state opacity and infinite-step opacity. For example, $\delta$-approximate initial-state opacity requires that, for any state run starting from a secret state, there exists another state run starting from a non-secret state, such that their corresponding output runs are $\delta$-close to each other. Intuitively, $\delta$-approximate initial-state opacity says that the intruder can never determine that the system is initiated from a secret state if it does not have an enough measurement precision which is captured by parameter $\delta$. In other words, instead of requiring that the system is exactly opaque, our new definitions essentially provide relaxed versions of opacity with a quantitative security guarantees level with respect to the measurement precision of the intruder.

For systems whose state-spaces are very large or even infinite, it is desirable to construct abstract models that preserve opacity, to some extent, for the purpose of verification. To this end, we propose the concept of $\varepsilon$-approximate opacity preserving simulation relation. We show that if there is an $\varepsilon$-approximate opacity preserving simulation relation from system $S_a$ to system $S_b$, then $S_b$ being $(\delta - 2\varepsilon)$-approximate opaque implies that $S_a$ is $\delta$-approximate opaque. In particular, for a class of incrementally input-to-state stable discrete-time control systems with possibly infinite state-spaces, we propose an effective approach to construct symbolic models (a.k.a. finite abstractions) that approximately simulate the original systems in the sense of opacity preserving and vice versa. The resulting symbolic model is finite if the state-space of the original continuous system is within a bounded region. Therefore, the proposed abstraction technique together with the verification algorithm for the finite case provide a sound way for verifying opacity of discrete-time control systems with continuous state-spaces.

The contributions of this work are summarized as follows.

- New simulation relations termed as $\varepsilon$-approximate opacity preserving simulation relations are proposed to characterize how close two systems are in terms of the satisfaction of approximate opacity.
- For a class of discrete-time control systems, we show how to construct symbolic models that preserve opacity with given a-priori precision.

### C. Related Works

Our work is closely related to several works in the literature. First, several different approaches have been proposed in the literature to evaluate opacity more quantitatively rather than requiring that the system is opaque exactly [14], [21]–[23]. For example, in [22], the authors adopt the Jensen-Shannon divergence as the measurement to quantify secrecy loss. In [14], [21], [23], stochastic DES models are used to study the probabilistic measurement of opacity. These approaches essentially aim to analyze how opaque a single system is, e.g., the probability of being opaque. However, they neither consider how close two systems are in terms of being opaque nor consider under what observation precision level, we can guarantee opacity.

There are also attempts in the literature to extend opacity from discrete systems to continuous systems. For example, in the recent results in [24]–[26], the authors extended the notion of opacity to (switched) linear systems. However, their definition of opacity is more related to an output reachability property rather than an information-flow property. Moreover, their formulation is mostly based on the setting of exact opacity, i.e., we can always distinguish two different outputs precisely no matter how close they are. In [24], the authors mentioned the direction of using output metric to quantify opacity and a property called strong $\epsilon$-$K$-initial-state opacity was proposed, which is closely related to our notions. However, no systematic study, e.g., verification and abstraction as we consider in this paper, was provided for this property.

Regarding the techniques used in this paper, first, our algorithms for the verification of approximate notions of opacity are motivated by the verification algorithms for exact opacity studied in [5], [27]. In particular, we use the idea of constructing a new system, called the state-estimator, that tracks all possible states consistent with the observation. However, our construction of state-estimator is not exactly the same as the existing one as additional state information is needed in order to handle the issue of approximation.

Abstraction-based techniques have also been investigated in the literature for the verification and synthesis of opacity; see, e.g., [28]–[32]. In particular, in our recent work [28], we propose several notions of opacity preserving (bi)simulation relations. However, these relations only preserve exact opacity for non-metric systems. Our new relations extend the relations in [28] to metric systems by taking into account how close two systems are. Such an extension is motivated by the definition of approximate (bi)simulation relation originally proposed in [33]. However, the original definition of approximate (bi)simulation relation does not necessarily preserve approximate opacity. Constructing symbolic models for control
systems is also an active research area; see, e.g., [34]–[37]. However, most of the existing works on the construction of symbolic models only consider the dynamics of the systems and are not taking into account the opacity property. In our approach, we need to consider both the dynamic and the secret of the system while constructing the symbolic model and guarantee the preservation of approximate opacity across related systems.

A related notion called differential privacy was introduced in [38] for database systems and has attracted significant attention in the past few years [39]–[41]. In particular, [40] extends the original notion of differential privacy to symbolic systems. Differential privacy requires that any two adjacent data should produce indistinguishable outputs in the probability sense. However, the essence of opacity is a confidentiality property that captures the plausible deniability of the system’s secret behavior, while differential privacy captures whether or not any sensitive data can be learned under some side-information. These two properties are incomparable in general. Note that there are also probabilistic versions of opacity studied in the literature for systems modeled as Markov chains [14], [21]–[23]. In those studies, the essence of probabilistic opacity is still plausible deniability but with a quantitative measure; the output at each state is still non-probabilistic.

Finally, approximate notions of two related properties called diagnosability and predictability are investigated recently in [42], [43]. Their setting is very similar to us as we both consider a measurement uncertainty threshold. However, diagnosability and predictability can be preserved by standard approximate simulation relation. We show that standard approximate simulation relation does not preserve opacity. Therefore, the proposed approximate opacity preserving simulation relation is different from the standard approximate simulation relation in the literature.

D. Organization

The rest of this paper is organized as follows. In Section II, we first introduce some necessary preliminaries. Then we propose the concept of approximate opacity in Section III. The verification procedures for approximate opacity are provided in Section IV. In Section V, approximate opacity preserving simulation relations are proposed and their properties are also discussed. In Section VI, we describe how to construct approximate opacity preserving symbolic models for incrementally stable discrete-time control systems with continuous state-spaces. Finally, we conclude the paper by Section VII. Preliminary and partial version of this paper is presented as an extended abstract in [44].

II. PRELIMINARIES

A. Notation

The symbols $\mathbb{N}$, $\mathbb{N}_0$, $\mathbb{Z}$, $\mathbb{R}$, $\mathbb{R}^+$, and $\mathbb{R}^+_0$ denote the set of natural, nonnegative integer, integer, real, positive, and non- negative real numbers, respectively. Given a vector $x \in \mathbb{R}^n$, we denote by $x_i$ the $i$-th element of $x$, and by $\|x\|$ the infinity norm of $x$. The closed ball centered at $u \in \mathbb{R}^m$ with radius $\lambda$ is defined by $B_\lambda(u) = \{v \in \mathbb{R}^m | \|u - v\| \leq \lambda\}$. We denote the closed ball centered at the origin in $\mathbb{R}^n$ and with radius $\lambda$ by $B_\lambda$. A set $B \subseteq \mathbb{R}^m$ is called a box if $B = \prod_{i=1}^m [c_i, d_i]$, where $c_i, d_i \in \mathbb{R}$ with $c_i < d_i$ for each $i \in \{1, \ldots, m\}$. The span of a box $B$ is defined as $\text{span}(B) = \min\{|d_i - c_i| \mid i = 1, \ldots, m\}$. For a box $B \subseteq \mathbb{R}^m$ and $\mu \leq \text{span}(B)$, define the $\mu$-approximation $[B]_\mu = \{a \in \mathbb{R}^m \mid a_i = k_i, k_i \in \mathbb{Z}, i = 1, \ldots, m\}$. Remark that $[B]_\mu \neq \emptyset$ for any $\mu \leq \text{span}(B)$. Geometrically, for any $\mu \in \mathbb{R}^+$ with $\mu \leq \text{span}(B)$ and $\lambda \geq \mu$, the collection of sets $\{[B_\lambda(p)]\}_{p \in [B]_\mu}$ is a finite covering of $B$, i.e., $B \subseteq \bigcup_{p \in [B]_\mu} B_\lambda(p)$. We extend the notions of span and approximation to finite unions of boxes as follows. Let $A = \bigcup_{j=1}^m A_j$, where each $A_j$ is a box. Define $\text{span}(A) = \min\{\text{span}(A_j) \mid j = 1, \ldots, M\}$, and for any $\mu \leq \text{span}(A)$, define $[A]_\mu = \bigcup_{j=1}^m [A_j]_\mu$. The Minkowski sum of two sets $P, Q \subseteq \mathbb{R}^n$ is defined by $P + Q = \{x \in \mathbb{R}^n | \exists p \in P, q \in Q, x = p + q\}$. Given a set $S \subseteq \mathbb{R}^n$ and a constant $\theta \in \mathbb{R}_{\geq 0}$, we define a new set $S^\theta = S + B_\theta$ as the inflated version of the set $S$.

Given a function $f : \mathbb{N}_0^+ \to \mathbb{R}^n$, the (essential) supremum of $f$ is denoted by $\|f\|_\infty = (\text{ess}\sup\{\|f(k)\| | k \geq 0\})$. A continuous function $\gamma : \mathbb{R}_0^+ \to \mathbb{R}_0^+$ is said to belong to class $\mathcal{K}$ if it is strictly increasing and $\gamma(0) = 0$; $\gamma$ is said to belong to class $\mathcal{K}_\infty$ if $\gamma \in \mathcal{K}$ and $\gamma(r) \to \infty$ as $r \to \infty$. A continuous function $\beta : \mathbb{R}_0^+ \times \mathbb{R}_0^+ \to \mathbb{R}_0^+$ is said to belong to class $\mathcal{KL}$ if, for each fixed $s$, the map $\beta(r, s)$ belongs to class $\mathcal{K}$ with respect to $r$ and, for each fixed nonzero $r$, the map $\beta(r, s)$ is decreasing with respect to $s$ and $\beta(r, s) \to 0$ as $s \to \infty$. We identify a relation $R \subseteq A \times B$ with the map $R : A \to 2^B$ defined by $b \in R(a)$ iff $(a, b) \in R$. Given a relation $R \subseteq A \times B$, $R^{-1}$ denotes the inverse relation defined by $R^{-1} = \{(b, a) \in B \times A : (a, b) \in R\}$.

B. System Model

In this paper, we employ a notion of “system” introduced in [45] as the underlying model of CPS describing both continuous-space and finite control systems.

Definition II.1. A system $S$ is a tuple

$$S = (X, X_0, U, \longrightarrow, Y, H),$$

where

- $X$ is a (possibly infinite) set of states;
- $X_0 \subseteq X$ is a (possibly infinite) set of initial states;
- $U$ is a (possibly infinite) set of inputs;
- $\longrightarrow \subseteq X \times U \times X$ is a transition relation;
- $Y$ is a set of outputs;
- $H : X \to Y$ is an output map.

A transition $(x, u, x') \in \longrightarrow$ is also denoted by $x \xrightarrow{u} x'$. For a transition $x \xrightarrow{u} x'$, state $x'$ is called a $u$-successor, or simply a successor, of state $x$; state $x$ is called a $u$-predecessor, or simply a predecessor, of state $x'$. We denote by $\text{Post}_u(x)$ the set of all $u$-successors of state $x$ and by $\text{Pre}_u(x)$ the set of all $u$-predecessors of state $x$. For a set of states $q \in 2^X$, we define $\text{Post}_u(q) = \bigcup_{x \in q} \text{Post}_u(x)$ and $\text{Pre}_u(q) = \bigcup_{x \in q} \text{Pre}_u(x)$. A system $S$ is said to be
metric, if the output set $Y$ is equipped with a metric $d: Y \times Y \to \mathbb{R}_+^+$;
finite (or symbolic), if $X$ and $U$ are finite sets;
deterministic, if for any state $x \in X$ and any input $u \in U$, $|\text{Post}_x(u)| \leq 1$ and nondeterministic otherwise.

Given a system $S = (X, X_0, U, \rightarrow, Y, H)$ and any initial state $x_0 \in X_0$, a finite state run generated from $x_0$ is a finite sequence of transitions:

$$x_0 \xrightarrow{u_1} x_1 \xrightarrow{u_2} \cdots \xrightarrow{u_{n-1}} x_{n-1} \xrightarrow{u_n} x_n,$$

such that $x_i \xrightarrow{u_{i+1}} x_{i+1}$ for all $0 \leq i < n$. A finite output run is a sequence $y_0 y_1 \ldots y_n$ such that there exists a finite state run of the form (2) with $y_i = H(x_i)$, for $i = 0, \ldots, n$.

### III. Exact and Approximate Opacity

In this section, we first review the notion of exact opacity. Then we introduce the notion of approximate opacity.

#### A. Exact Opacity

In many applications, systems may have some “secrets” that do not want to be revealed to intruders that are potentially malicious. In this paper, we adopt a state-based formulation of secrets. Specifically, we assume that $X_S \subseteq X$ is a set of secret states. Hereafter, we will always consider systems with secret states and we write a system $S = (X, X_0, U, \rightarrow, Y, H)$ with secret states $X_S$ by a new tuple $S = (X, X_0, X_S, U, \rightarrow, Y, H)$.

In order to characterize whether or not a system is secure, the concept of opacity was proposed in the literature. We review three basic notions of opacity [9] as follows.

**Definition III.1.** Consider a system $S = (X, X_0, X_S, U, \rightarrow, Y, H)$. System $S$ is said to be

- **initial-state opaque** if for any $x_0 \in X_0 \cap X_S$ and finite state run $r_0 \xrightarrow{u_1} x_1 \xrightarrow{u_2} \cdots \xrightarrow{u_n} x_n$, there exist $x_0' \in X_0 \setminus X_S$ and a finite state run $r_0' \xrightarrow{u_1'} x_1' \xrightarrow{u_2'} \cdots \xrightarrow{u_n'} x_n'$ such that $H(x_i) = H(x_i')$ for any $i = 0, 1, \ldots, n$;
- **current-state opaque** if for any $x_0 \in X_0$ and finite state run $r_0 \xrightarrow{u_1} x_1 \xrightarrow{u_2} \cdots \xrightarrow{u_n} x_n$ such that $x_n \in X_S$, there exist $x_0' \in X_0$ and finite state run $r_0' \xrightarrow{u_1'} x_1' \xrightarrow{u_2'} \cdots \xrightarrow{u_n'} x_n'$ such that $x_i' \in X \setminus X_S$ and $H(x_i) = H(x_i')$ for any $i = 0, 1, \ldots, n$;
- **infinite-step opaque** if for any $x_0 \in X_0$, any finite state run $r_0 \xrightarrow{u_1} x_1 \xrightarrow{u_2} \cdots \xrightarrow{u_n} x_n$ and any $k \in \{0, \ldots, n\}$, $x_k \in X_S$ implies that there exist $x_0' \in X_0$ and a finite state run $r_0' \xrightarrow{u_1'} x_1' \xrightarrow{u_2'} \cdots \xrightarrow{u_n'} x_n'$ such that $x_i' \in X \setminus X_S$ and $H(x_i) = H(x_i')$ for any $i = 0, 1, \ldots, n$.

The intuitions of the above definitions are as follows. Suppose that the output run of the system can be observed by a passive intruder that may use this information to infer the secret of the system. Then initial-state opacity requires that the intruder should never know for sure that the system is initiated from a secret state no matter what output run is generated. Similarly, current-state opacity says that the intruder should never know for sure that the system is currently at a secret state no matter what output run is generated. Infinite-step opacity is stronger than both initial-state opacity and current-state opacity as it requires that the intruder should never know that the system is/was at a secret state for any specific instant $k$. For any system $S = (X, X_0, X_S, U, \rightarrow, Y, H)$, we assume without loss of generality that $\forall x_0 \in X_0 : \{ x \in X_0 : H(x) = (H(x_0)) \subseteq X_S$. This assumption essentially requires that the secret of the system cannot be revealed initially; otherwise, the system is not opaque trivially.

**Remark III.2.** Definition III.1 implicitly considers the following model of the intruder: (i) the intruder knows the model of the system; and (ii) it can only observe the output trajectory of the system. Therefore, the intruder essentially wants to use the output trajectory observed online and the knowledge of the system model to infer the internal behavior/state of the system. Note that, in our setting, the input information is assumed to be internal and the intruder does not know which input the system takes. This setting can be easily relaxed and all results in this paper can be extended to the case where both input and output information are available by the intruder. For example, we can simply refine the model of the system such that the output space of the refined system is a pair and the input leading to a state is also encoded in the output of this state.

**Remark III.3.** Our definition of infinite-step opacity requires that the intruder should never know for sure that the system is/was at a secret state for any specific instant. In some cases, the intruder may know that the system must have visited a secret state, although it cannot tell the precise instant. Such a requirement can be captured by the notion of strong (or trajectory-based) infinite-step opacity; see, e.g., Remark 5 in [6]. This definition is stronger than ours and which one to use is dependent on the applications. However, strong infinite-step opacity can be transformed to current-state opacity by augmenting the state-space to encode whether a secret state has been visited or not.

#### B. Approximate Opacity

Note that Definition III.1 requires that for any secret behavior, there exists a non-secret behavior such that they generate exactly the same output. Therefore, we will also refer to these definitions as exact opacity. Exact opacity essentially assumes that the intruder or the observer can always measure each output or distinguish between two different outputs precisely. This setting is reasonable for non-metric systems where outputs are symbols or events. However, for metric systems, e.g., when the outputs are physical signals, this setting may be too restrictive. In particular, due to the imperfect measurement precision, which is almost the case for all physical systems, it is very difficult to distinguish two observations if their difference is very small. Therefore, exact opacity may be too strong for metric systems and it will be useful to define a weak and “robust” version of opacity by characterizing under which measurement precision the system is opaque. To this end, we define new notions of opacity called approximate opacity for metric systems.
Definition III.4. Let $S = (X, X_0, X_S, U, \rightarrow, Y, H)$ be a metric system, with the metric $d$ defined over the output set, and a constant $\delta \geq 0$. System $S$ is said to be
- $\delta$-approximate initial-state opaque if for any $x_0 \in X_0 \cap X_S$ and finite state run $x_0 \xrightarrow{u_1} x_1 \xrightarrow{u_2} \cdots \xrightarrow{u_n} x_n$, there exist $x'_0 \in X_0 \setminus X_S$ and a finite state run $x'_0 \xrightarrow{u'_1} x'_1 \xrightarrow{u'_2} \cdots \xrightarrow{u'_n} x'_n$ such that
$$\max_{i \in \{0, \ldots, n\}} d(H(x_i), H(x'_i)) \leq \delta$$
- $\delta$-approximate current-state opaque if for any $x_0 \in X_0$ and finite state run $x_0 \xrightarrow{u_1} x_1 \xrightarrow{u_2} \cdots \xrightarrow{u_n} x_n$ such that $x_n \in X_S$, there exist $x'_0 \in X_0$ and finite state run $x'_0 \xrightarrow{u'_1} x'_1 \xrightarrow{u'_2} \cdots \xrightarrow{u'_n} x'_n$ such that $x'_n \in X \setminus X_S$ and
$$\max_{i \in \{0, \ldots, n\}} d(H(x_i), H(x'_i)) \leq \delta$$
- $\delta$-approximate infinite-step opaque if for any $x_0 \in X_0$, any finite state run $x_0 \xrightarrow{u_1} x_1 \xrightarrow{u_2} \cdots \xrightarrow{u_n} x_n$ and any $k \in \{0, \ldots, n\}$, $x_k \in X_S$ implies that there exist $x'_0 \in X_0$ and a finite state run $x'_0 \xrightarrow{u'_1} x'_1 \xrightarrow{u'_2} \cdots \xrightarrow{u'_n} x'_n$ such that $x'_k \in X \setminus X_S$ and
$$\max_{i \in \{0, \ldots, n\}} d(H(x_i), H(x'_i)) \leq \delta.$$

The notions of $\delta$-approximate initial-state, current-state opacity and infinite-step opacity are very similar to their exact counterparts. The main difference is how we treat two outputs as indistinguishable outputs. Specifically, same as the exact case, we still assume that the intruder know the system model and the output trajectory generated. However, we further assume that the intruder may not be able to distinguish an output trajectory from other $\delta$-closed trajectories confidentially. Intuitively, the approximate version of opacity can be interpreted as “the secret of the system cannot be revealed to an intruder that does not have an enough measurement precision related to parameter $\delta$”. In other words, instead of providing an exact security guarantee, approximate opacity provides a relaxed and quantitative security guarantee with respect to the measurement precision of the intruder. Therefore, the value $\delta$ can be interpreted as either the measurement imprision of the intruder or the security level the system can guarantee, i.e., under how powerful intruder the system is still secure. Clearly, when $\delta = 0$, each notion of $\delta$-approximate opacity reduces to its exact version. Similar to the exact case, hereafter, we assume without loss of generality that
$$\forall x_0 \in X_0 : \{x \in X_0 : d(H(x_0), H(x)) \leq \delta\} \not\subseteq X_S,$$
for any system $S = (X, X_0, X_S, U, \rightarrow, Y, H)$. This assumption can be easily checked and its non-satisfaction means that $\delta$-approximate initial-state opacity, $\delta$-approximate current-state opacity and $\delta$-approximate infinite-step opacity are all violated trivially.

We illustrate exact opacity and approximate opacity by the following example.

Example III.5. Consider system $S = (X, X_0, X_S, U, \rightarrow, Y, H)$ depicted in Figure 1, where
A. Verification of Approximate Initial-State Opacity

In order to verify $\delta$-approximate initial-state opacity, we construct a new system called the $\delta$-approximate initial-state estimator defined as follows.

**Definition IV.1.** Let $S = (X, X_0, X_S, U, Y, H)$ be a metric system, with the metric $d$ defined over the output set, and a constant $\delta \geq 0$. The $\delta$-approximate initial-state estimator is a system (without outputs)

$$S_I = (X_I, X_{I0}, U, Y),$$

where

- $X_I \subseteq X \times 2^X$ is the set of states;
- $X_{I0} = \{(x, q) \in X \times 2^X : x' \in q \iff d(H(x), H(x')) \leq \delta\}$ is the set of initial states;
- $U$ is the set of inputs, which is the same as the one in $S$;
- $I \subseteq X_I \times U \times X_I$ is the transition function defined by: for any $(x, q), (x', q') \in X \times 2^X$ and $u \in U$,
- $(x, q) \xrightarrow{u} (x', q')$ if
  1. $(x', u, x) \in I$; and
  2. $q' = \cup_{u \in U} \text{Pre}_u(q) \cap \{x'' \in X : d(H(x'), H(x'')) \leq \delta\}$.

For the sake of simplicity, we only consider the part of $S_I$ that is reachable from initial states.

Intuitively, the $\delta$-approximate initial-state estimator works as follows. Each initial state of $S_I$ is a pair consisting of a system state and its $\delta$-closed states; we consider all such pairs as the set of initial states. Then from each state, we track backwards states that are consistent with the output information recursively. Our construction is motivated by the reversed-automaton-based initial-state-estimator proposed in [9] but with the following differences. First, the way we defined information-consistency is different. Here we treat states whose output are $\delta$-close to each other as consistent states. Moreover, the structure in [9] requires only a space state of $2^X$, while our state space is $X \times 2^X$. The additional first component can be understood as the “reference trajectory” that is used to determine what is “$\delta$-close” at each instant. We use the following result to show the main property of $S_I$.

**Proposition IV.2.** Let $S = (X, X_0, X_S, U, Y, H)$ be a metric system, with the metric $d$ defined over the output set, and a constant $\delta \geq 0$. Let $S_I = (X_I, X_{I0}, U, Y)$ be its $\delta$-approximate initial-state estimator. Then for any $(x_0, q_0) \in X_{I0}$ and any finite run

$$(x_0, q_0) \xrightarrow{u_1} (x_1, q_1) \xrightarrow{u_2} \cdots \xrightarrow{u_n} (x_n, q_n)$$

we have

(i) $x_n = \cup_{i=1}^n x_{i-1}$ and

(ii) $q_n = \{x' \in X : \exists x' \xrightarrow{u_i} x_{i-1} \xrightarrow{u_{i-1}} \cdots x_0; \max_{i \in [1, \ldots, n]} d(H(x_i), H(x'_{i-1})) \leq \delta\}$.

Proof. See the Appendix.

The next theorem provides one of the main results of this section on the verification of $\delta$-approximate initial-state opacity of finite metric systems.

**Theorem IV.3.** Let $S = (X, X_0, X_S, U, Y, H)$ be a metric system, with the metric $d$ defined over the output set, and a constant $\delta \geq 0$. Let $S_I = (X_I, X_{I0}, U, Y)$ be its $\delta$-approximate initial-state estimator. Then, $S$ is $\delta$-approximate initial-state opaque if and only if

$$\forall (x, q) \in X_I : x \in X_0 \cap X_S \Rightarrow q \cap X_0 \subseteq X_S. \tag{3}$$

Proof. See the Appendix.

We illustrate how to verify $\delta$-approximate initial-state opacity by the following example.

**Example IV.4.** Let us still consider system $S$ shown in Figure 1. The $\delta$-approximate initial-state estimator $S_I$ when $\delta = 0.1$ is shown in Figure 2(a). For example, for initial state $(D, \{D\})$, we have $(D, \{D\}) \xrightarrow{u} (B, \{B, C\})$ since $B \xrightarrow{D}$ $D$ and $\{B, C\} = \text{Pre}_u(D) \cap \{x \in X : d(0.1, H(x)) \leq 0.1\} = \{B, C\} \cap \{B, A, B, C\}$. However, for state $(B, \{B, C\}) \in X_I$, we have $B \cup X_0 \cap X_S$ and $\{B, C\} \cap X_0 = \{B\} \subseteq X_S$. Therefore, by Theorem IV.3, we know that the system is not $0.1$-approximate initial-state opaque. Similarly, we can also construct $S_I$ for the case of $\delta = 0.15$, which is shown in Figure 2(b). Since for state $(B, \{B, A, C\}) \in X_I$, which is the only state whose first component is in $X_0 \cap X_S$, we have $\{B, A, C\} \cap X_0 = \{A, B\} \subseteq X_S$. By Theorem IV.3, we know that the system is $0.15$-approximate initial-state opaque.

B. Verification of Approximate Current-State Opacity

In order to verify $\delta$-approximate current-state opacity, we also need to construct a new system called the $\delta$-approximate current-state estimator defined as follows.

**Definition IV.5.** Let $S = (X, X_0, X_S, U, Y, H)$ be a metric system, with the metric $d$ defined over the output set.

![Diagram](image-url)
set, and a constant $\delta \geq 0$. The $\delta$-approximate current-state estimator is a system (without outputs)

$$S_C = (X_C, X_{C0}, U, \xrightarrow{C})$$

where

- $X_C \subseteq X \times 2^X$ is the set of states;
- $X_{C0} = \{(x, q) \in X_0 \times 2^{X_0} : x' \in q \iff d(H(x), H(x')) \leq \delta\}$ is the set of initial states;
- $U$ is the set of inputs, which is the same as the one in $S$;
- $\xrightarrow{C} \subseteq X \times U \times X_C$ is the transition function defined by: for any $(x, q, (x', q')) \in X \times 2^X$ and $u \in U$,

$$(x, q) \xrightarrow{u} (x', q')$$

if

1. $(x, u, x') \in U$; and
2. $q' = \cup_{u \in U} \text{Post}_u(x') \cap \{x'' \in X : d(H(x'), H(x'')) \leq \delta\}$.

For the sake of simplicity, we only consider the part of $S_C$ that is reachable from initial states.

The construction of $S_C$ is similar to $S_I$. However, we need to track all forward runs from each pair of initial-state and its information-consistent states. Still, we need the first component as the "reference state" to determine what are "$\delta$-close" states. We use the following result to state the main properties of $S_C$.

**Proposition IV.6.** Let $S = (X, X_0, X_S, U, \rightarrow, Y, H)$ be a metric system, with the metric $d$ defined over the output set, and a constant $\delta \geq 0$. Let $S_C = (X_C, X_{C0}, U, \xrightarrow{C})$ be its $\delta$-approximate current-state estimator. Then for any $(x_0, q_0) \in X_{C0}$ and any finite run

$$(x_0, q_0) \xrightarrow{u_1} (x_1, q_1) \xrightarrow{u_2} \cdots \xrightarrow{u_n} (x_n, q_n),$$

we have

(i) $x_0 \xrightarrow{u_1} x_1 \xrightarrow{u_2} \cdots \xrightarrow{u_n} x_n$; and
(ii) $q_0 = \{x_n \in X : \exists x'_0 \in X_0, \exists x'_0 \xrightarrow{u_1} x'_1 \xrightarrow{u_2} \cdots \xrightarrow{u_n} x'_n \text{ s.t. } \max_{i \in [0, 1, \ldots, n]} d(H(x_i), H(x'_i)) \leq \delta\}$.  

Proof. See the Appendix.

Now, we show the second main result of this section by providing a verification scheme for $\delta$-approximate current-state opacity of finite metric systems.

**Theorem IV.7.** Let $S = (X, X_0, X_S, U, \rightarrow, Y, H)$ be a metric system, with the metric $d$ defined over the output set, and a constant $\delta \geq 0$. Let $S_C = (X_C, X_{C0}, U, \xrightarrow{C})$ be its $\delta$-approximate current-state estimator. Then, $S$ is $\delta$-approximate current-state opaque if and only if

$$\forall (x, q) \in X_C : q \not\subseteq X_S.$$  

(4)

Proof. See the Appendix.

**C. Verification of Approximate Infinite-Step Opacity**

Finally, we can combine the $\delta$-approximate initial-state estimator $S_I$ and the $\delta$-approximate current-state estimator $S_C$ to verify $\delta$-approximate infinite-step opacity of finite metric systems. The verification scheme is provided by the following theorem.

**Theorem IV.8.** Let $S = (X, X_0, X_S, U, \rightarrow, Y, H)$ be a finite metric system, with the metric $d$ defined over the output set, and a constant $\delta \geq 0$. Let $S_I = (X_I, X_{I0}, U, \xrightarrow{I})$ and $S_C = (X_C, X_{C0}, U, \xrightarrow{C})$ be its $\delta$-approximate initial-state estimator and $\delta$-approximate current-state estimator, respectively. Then, $S$ is $\delta$-approximate infinite-step opaque if and only if

$$\forall (x, q) \in X_I, (x', q') \in X_C : x = x' \in X_S \Rightarrow q \cap q' \not\subseteq X_S.$$  

(5)

Proof. See the Appendix.

**Remark IV.9.** We conclude this section by discussing the complexity of verifying approximate opacity. Let $S = (X, X_0, X_S, U, \rightarrow, Y, H)$ be a finite metric system. The complexity of the verification algorithms for both approximate initial-state and current-state opacity is $O(|U| \times |X| \times 2^{|X|})$, which is the size of $S_I$ or $S_C$. For approximate infinite-step opacity, we need to construct both $S_I$ and $S_C$, and compare each pair of states in $S_I$ and $S_C$. Therefore, the complexity for verifying approximate infinite-step opacity using Theorem IV.8 is $O(|U| \times |X|^2 \times 4^{|X|})$. It is worth noting that the complexity of verifying exact opacity as in Definition III.1 is already known to be PSPACE-complete [46]. Using a similar reduction, we can conclude that the complexity of verifying approximate opacity as in Definition III.4 is also PSPACE-complete for $\delta > 0$. Finally, we note that the exponential complexity essentially comes from the subset construction to handle information uncertainty. In practice, the subset construction usually results in a quite small structure; see, e.g., [47] for detailed empirical studies on this issue.

V. APPROXIMATE SIMULATION RELATIONS FOR OPACITY

In the previous sections, we have introduced notions of approximate opacity and their verification procedures. However, when the system is very large or even infinite, verifying opacity based on the original system is not efficient or not even possible. Therefore, it will be beneficial if we can verify opacity based on an “equivalent” smaller or symbolic system. To this end, in this section, we study under what conditions two systems are equivalent and in what sense. Specifically, we introduce new notions of approximate opacity preserving simulation relations, inspired by the one in [33]. The newly proposed simulation relations will provide the basis for abstraction-based verification of approximate opacity.

**A. Approximate Initial-State Opacity Preserving Simulation Relation**

First, we introduce a new notion of approximate initial-state opacity preserving simulation relation.

**Definition V.1.** (Approximate Initial-State Transparency Preserving Simulation Relation) Consider two metric systems $S_a = (X_a, X_{a0}, X_{aS}, U_a, \rightarrow, Y_a, H_a)$ and $S_b = (X_b, X_{b0}, X_{bS}, U_b, \rightarrow, Y_b, H_b)$ with the same output sets $Y_a = Y_b$ and metric $d$. For $\epsilon \in \mathbb{R}^+$, a relation $R \subseteq X_a \times X_b$ is called an $\epsilon$-approximate initial-state transparency preserving
The following corollary is a simple consequence of the result in Theorem V.2 but for the lack of \( \delta \)-approximate initial-state opacity.

**Corollary V.3.** Let \( S_a = (X_a, X_{a0}, X_{aS}, U_a, \rightarrow_a, Y_a, H_a) \) and \( S_b = (X_b, X_{b0}, X_{bS}, U_b, \rightarrow_b, Y_b, H_b) \) be two metric systems with the same output sets \( Y_a = Y_b \) and metric \( d \) and let \( \varepsilon, \delta \in \mathbb{R}^{+}_{0} \). If \( S_b \preceq_{\varepsilon}^2 S_a \), then the following implication hold:

\[
S_b \text{ is not } (\delta + 2\varepsilon)\text{-approximate initial-state opaque} \\
\Rightarrow S_a \text{ is not } \delta\text{-approximate initial-state opaque.}
\]

**Proof.** Since \( S_b \preceq_{\varepsilon}^2 S_a \), by Theorem V.2, we know that \( S_a \) being \( \delta \)-approximate initial-state opaque implies that \( S_b \) is \( (\delta + 2\varepsilon)\)-approximate initial-state opaque. Hence, \( S_a \) not being \( (\delta + 2\varepsilon)\)-approximate initial-state opaque implies that \( S_a \) is not \( \delta \)-approximate initial-state opaque.

**Remark V.4.** It is worth remarking that \( \delta \) and \( \varepsilon \) are parameters specifying two different types of precision. Parameter \( \delta \) is used to specify the measurement precision under which we can guarantee opacity for a single system, while parameter \( \varepsilon \) is used to characterize the “distance” between two systems in terms of being approximate opaque. The reader should not be confused by the different roles of these two parameters.

We illustrate \( \varepsilon \)-approximate initial-state opacity preserving simulation relation by the following example.

**Example V.5.** Let us consider systems \( S_a \) and \( S_b \) shown in Figures 3(a) and 3(b), respectively. We mark all secret states by red and the output map is specified by the value associated to each state. Let us consider the following relation \( R = \{(A, J), (B, K), (C, K), (D, K), (E, N), (F, M), (G, M), (I, M)\} \). We claim that \( R \) is an \( \varepsilon \)-approximate initial-state opacity preserving simulation relation from \( S_a \) to \( S_b \) when \( \varepsilon = 0.1 \). We check item by item following Definition V.1. First, for \( E \in X_a \cap X_{aS} \), we have \( N \in X_{b0} \cap X_{bS} \) such that \( (E, N) \in R \). Similarly, for \( J \in X_{b0} \cap X_{bS} \), we have \( A \in X_{a0} \cap X_{aS} \) such that \( (A, J) \in R \). Therefore, condition 1) in Definition V.1 holds. Also, for any \( (x_a, x_b) \in R \), we have \( d(H_a(x_a), H_b(x_b)) \leq 0.1 \), e.g., \( d(H_a(A), H_b(J)) = 0.1 \) and \( d(H_a(C), H_b(K)) = 0 \). Therefore, condition 2) in Definition V.1 holds. Finally, we can also check that condition 3) in Definition V.1 holds. For example, for \( (D, K) \in R \) and \( D \rightarrow_a B \), we can choose \( K \rightarrow_b K \) such that \( (B, K) \in R \); for \( (E, M) \in R \) and \( N \rightarrow_b M \), we can choose \( E \rightarrow_b F \) such that \( (F, M) \in R \). Therefore, we know that \( R \) is an \( \varepsilon \)-InitSOP simulation relation from \( S_a \) to \( S_b \), i.e., \( S_a \preceq_{\varepsilon}^2 S_b \).

Then, by applying the verification algorithm in Section IV, we can check that \( S_b \) is \( \delta \)-approximate initial-state opaque for \( \delta = 0.1 \). Therefore, according to Theorem V.2, we conclude that \( S_a \) is \( 0.3 \)-approximate initial-state opaque, where \( 0.3 = \delta + 2\varepsilon \), without applying the verification algorithm to \( S_a \) directly.
We say that

\[\mathcal{S}_a \subseteq \mathcal{S}_b,\]

by conditions 1), 2), 3)-a) and 3)-b) in Definition V.6., there exist an initial state \(x_0 \in X_0\) and a finite run \(x'_0 \xrightarrow{u'_1} x'_1 \xrightarrow{u'_2} \cdots \xrightarrow{u'_n} x'_n\) in \(\mathcal{S}_b\) such that \(x'_n \in X_{b, S}\) and

\[\forall i \in \{0, 1, \ldots, n\}: d(H_a(x_i), H_b(x'_i)) \leq \varepsilon.\]

(10)

Since \(\mathcal{S}_b\) is \((\delta - 2\varepsilon)\)-approximate current-state opaque, there exist an initial state \(x''_0 \in X_0\) and a finite run \(x''_0 \xrightarrow{u''_1} x''_1 \xrightarrow{u''_2} \cdots \xrightarrow{u''_n} x''_n\) such that \(x''_n \in X_0 \setminus X_{b, S}\) and

\[\max_{i \in \{0, 1, \ldots, n\}} d(H_b(x''_i), H_b(x''_i)) \leq \delta - 2\varepsilon.\]

(11)

Again, since \(\mathcal{S}_a \subseteq \mathcal{S}_b,\) by conditions 1), 2), 3)-c) and 3)-d) in Definition V.6., there exist an initial state \(x'''_0 \in X_0\) and a finite run \(x'''_0 \xrightarrow{u'''_1} x'''_1 \xrightarrow{u'''_2} \cdots \xrightarrow{u'''_n} x'''_n\) such that \(x'''_n \in X_0 \setminus X_{a, S}\) and

\[\forall i \in \{0, 1, \ldots, n\}: d(H_a(x'''_i), H_b(x''''_i)) \leq \varepsilon.\]

(12)

Combining equations (10), (11), (12), and using the triangle inequality, we have

\[\max_{i \in \{0, 1, \ldots, n\}} d(H_a(x_i), H_a(x'_i)) \leq \delta.\]

(13)

Since \(x_0 \in X_0\) and \(x_0 \xrightarrow{u_1} x_1 \xrightarrow{u_2} \cdots \xrightarrow{u_n} x_n\) are arbitrary, we conclude that \(\mathcal{S}_a\) is \(\delta\)-approximate current-state opaque.

\[\square\]

C. Approximate Infinite-Step Opacity Preserving Simulation Relation

Finally, by combining \(\varepsilon\)-CurSOP simulation relation and \(\varepsilon\)-InitSOP simulation relation, we provide a notion of approximate simulation relation for preserving infinite-step opacity.

Definition V.8. (Approximate Infinite-Step Opacity Preserving Simulation Relation) Let \(\mathcal{S}_a = (X_a, X_{a, 0}, X_{a, S}, U_a, \triangleright \leftarrow \mapsto, Y_a, H_a)\) and \(\mathcal{S}_b = (X_b, X_{b, 0}, X_{b, S}, U_b, \triangleright \leftarrow \mapsto, Y_b, H_b)\) be two metric systems with the same output sets \(Y_a = Y_b\) and metric \(d\). For \(\varepsilon \in \mathbb{R}_+^+,\) a relation \(R \subseteq X_a \times X_b\) is called an \(\varepsilon\)-approximate infinite-step opacity preserving simulation relation (\(\varepsilon\)-InfSOP simulation relation) from \(\mathcal{S}_a\) to \(\mathcal{S}_b\) if it is both an \(\varepsilon\)-CurSOP simulation relation from \(\mathcal{S}_a\) to \(\mathcal{S}_b\) and an \(\varepsilon\)-InitSOP simulation relation from \(\mathcal{S}_a\) to \(\mathcal{S}_b\), i.e.,

1) a) \(\forall x_0 \in X_0, \exists x_0 \in X_{b, 0}: (x_0, x_0) \in R;\)

b) \(\forall x_0 \in X_0 \cap X_{a, S}, \exists x_0 \in X_{b, 0} \cap X_{b, S}: (x_0, x_0) \in R;\)

c) \(\forall x_0 \in X_{a, 0} \setminus X_{b, S}, \exists x_0 \in X_{b, 0} \setminus X_{b, S}: (x_0, x_0) \in R;\)

2) \(\forall (x_a, x_b) \in R: d(H_a(x_a), H_b(x_b)) \leq \varepsilon;\)

3) For any \((x_a, x_b) \in R,\) we have

a) \(\exists x' \xrightarrow{u'_1} x'_1 \xrightarrow{u'_2} \cdots \xrightarrow{u'_n} x'_n \in S_b\) such that \(d(H_a(x_0), H_b(x'_0)) \leq \delta.\)

b) \(\exists x' \xrightarrow{u''_1} x''_1 \xrightarrow{u''_2} \cdots \xrightarrow{u''_n} x''_n \in S_b\) such that \(d(H_a(x''_0), H_b(x_0)) \leq \varepsilon.\)

c) \(\exists x' \xrightarrow{u'_1} x'_1 \xrightarrow{u'_2} \cdots \xrightarrow{u'_n} x'_n \in S_b\) such that \(d(H_a(x_0), H_b(x'_0)) \leq \delta.\)

(14)

Proof.

Let us consider an arbitrary initial state \(x_0 \in X_0\) and a finite run \(x_0 \xrightarrow{u_1} x_1 \xrightarrow{u_2} \cdots \xrightarrow{u_n} x_n\) in \(\mathcal{S}_a\) such that \(x_n \in X_{a, S}\). Since we assume that \(\{x \in X_0 : (H_a(x_0), H_a(x)) \leq \delta\} \subseteq X_{a, S}\), we observe immediately that there exists \(x'_0 \in X_0 \setminus X_{a, S}\) such that \(d(H_a(x_0), H_b(x'_0)) \leq \delta)\). Then, we consider the case of \(n \geq 1\). Since \(\mathcal{S}_a \subseteq \mathcal{S}_b,\) by conditions 1), 2), 3)-a) and 3)-b) in Definition V.6., there exist an initial state \(x_0 \in X_0\) and a finite run \(x'_0 \xrightarrow{u'_1} x'_1 \xrightarrow{u'_2} \cdots \xrightarrow{u'_n} x'_n\) in \(\mathcal{S}_b\) such that \(x'_n \in X_{b, S}\) and

\[\forall i \in \{0, 1, \ldots, n\}: d(H_a(x_i), H_b(x'_i)) \leq \varepsilon.\]

(10)

Since \(\mathcal{S}_b\) is \((\delta - 2\varepsilon)\)-approximate current-state opaque, there exist an initial state \(x''_0 \in X_0\) and a finite run \(x''_0 \xrightarrow{u''_1} x''_1 \xrightarrow{u''_2} \cdots \xrightarrow{u''_n} x''_n\) such that \(x''_n \in X_0 \setminus X_{b, S}\) and

\[\max_{i \in \{0, 1, \ldots, n\}} d(H_b(x''_i), H_b(x''_i)) \leq \delta - 2\varepsilon.\]

(11)

Again, since \(\mathcal{S}_a \subseteq \mathcal{S}_b,\) by conditions 1), 2), 3)-c) and 3)-d) in Definition V.6., there exist an initial state \(x'''_0 \in X_0\) and a finite run \(x'''_0 \xrightarrow{u'''_1} x'''_1 \xrightarrow{u'''_2} \cdots \xrightarrow{u'''_n} x'''_n\) such that \(x'''_n \in X_0 \setminus X_{a, S}\) and

\[\forall i \in \{0, 1, \ldots, n\}: d(H_a(x'''_i), H_b(x''''_i)) \leq \varepsilon.\]

(12)

Combining equations (10), (11), (12), and using the triangle inequality, we have

\[\max_{i \in \{0, 1, \ldots, n\}} d(H_a(x_i), H_a(x'_i)) \leq \delta.\]

(13)

Since \(x_0 \in X_0\) and \(x_0 \xrightarrow{u_1} x_1 \xrightarrow{u_2} \cdots \xrightarrow{u_n} x_n\) are arbitrary, we conclude that \(\mathcal{S}_a\) is \(\delta\)-approximate current-state opaque.

\[\square\]
We say that $S_a$ is $\varepsilon$-InfSOP simulated by $S_b$, denoted by $S_a \leq_{IF}^\varepsilon S_b$, if there exists an $\varepsilon$-InfSOP simulation relation $R$ from $S_a$ to $S_b$.

Similar to the cases of initial-state opacity and current-state opacity, we have the following theorem as a sufficient condition for $\delta$-approximate infinite-step opacity based on related systems as in Definition V.8.

**Theorem V.9.** Let $S_a = (X_a, X_{a0}, X_{aS}, U_a, \rightarrow_a, Y_a, H_a)$ and $S_b = (X_b, X_{b0}, X_{bS}, U_b, \rightarrow_b, Y_b, H_b)$ be two metric systems with the same output sets $Y_a = Y_b$ and metric $d$ and let $\varepsilon, \delta \in \mathbb{R}_+^*$. If $S_a \leq_{IF}^\varepsilon S_b$ and $\varepsilon \leq \frac{\delta}{2}$, then the following implication holds:

$S_b$ is $(\delta - 2\varepsilon)$-approximate infinite-step opaque

$\Rightarrow S_a$ is $\delta$-approximate infinite-step opaque.

**Proof.** Let us consider an arbitrary initial state $x_0 \in X_{a0}$ and finite run $x_0 \xrightarrow{u^i} x_1 \xrightarrow{u^2} \cdots \xrightarrow{u^n} x_n$ in $S_a$ such that $x_k \in X_{aS}$ for some $k = 0, \ldots, n$. We consider the following two cases:

If $k = 0$, then we have $x_0 \in X_{aS}$. Since $S_a \leq_{IF}^\varepsilon S_b$, implies $S_a \leq_{IF}^\varepsilon S_b$, by the proof of Theorem V.2, we know that there exists an initial state $x_0^I \in X_{a0} \setminus X_{aS}$ and a run $x_0^I \xrightarrow{u^i} x_1^I \xrightarrow{u^2} \cdots \xrightarrow{u^n} x_n^I$ such that $\max_{i\in\{0,1,\ldots,n\}} d(H_a(x_i), H_b(x_i')) \leq \delta$.

If $k \geq 1$, then similar to the proof of Theorem V.7, by conditions 1)-(a), 2), 3)-(a), 3)-(b), 3)-(c) and 3)-(d) in Definition V.8 and the fact the $S_i$ is $(\delta - 2\varepsilon)$-approximate infinite-step opaque, there exist an initial state $x_0^I \in X_{a0}$ and a finite run $x_0^I \xrightarrow{u^i} x_1^I \xrightarrow{u^2} \cdots \xrightarrow{u^n} x_n^I$ such that $x_k' \in X_a \setminus X_{aS}$ and $\max_{i\in\{0,1,\ldots,n\}} d(H_a(x_i), H_b(x_i')) \leq \delta$.

Since $x_0 \in X_{a0}$, $x_0 \xrightarrow{u^i} x_1 \xrightarrow{u^2} \cdots \xrightarrow{u^n} x_n$ and index $k$ are arbitrary, we conclude that $S_a$ is $\delta$-approximate infinite-step opaque.

VI. OPACITY OF CONTROL SYSTEMS

In the previous section, we have introduced notions of approximate opacity-preserving simulation relation and discussed their properties. This allows us to verify approximate opacity for infinite systems, e.g., continuous dynamic systems, based on their finite abstractions. Then the following question arises naturally: how can we construct such an abstraction for a given system possibly with infinite number of states. In general, how to construct symbolic abstractions are system-dependent and not all systems admit symbolic models. In this section, we show that a class of discrete-time control systems do admit symbolic models for the purpose of verifying approximate opacity under certain stability assumption.

To be more specific, we consider a class of discrete-time control systems of the following form.

**Definition VI.1.** A discrete-time control system $\Sigma$ is defined by the tuple $\Sigma = (X, S, U, f, Y, h)$, where $X$, $S$, and $Y$ are the state, input, and output sets, respectively, and are subsets of normed vector spaces with appropriate finite dimensions. Set $S \subseteq X$ is a set of secret states. The map $f : X \times U \to X$ is called the transition function, and $h : X \to Y$ is the output map and assumed to satisfy the following Lipschitz condition:

$$\|h(x) - h(y)\| \leq \alpha(\|x - y\|) \text{ for some } \alpha \in K_\infty \text{ and all } x, y \in X.$$  

The discrete-time control system $\Sigma$ is described by difference equations of the form

$$\begin{align*}
\xi(k + 1) &= f(\xi(k), u(k)), \\
h(k) &= h(\xi(k)),
\end{align*}$$

where $\xi : N_0 \to X$, $\zeta : N_0 \to Y$, and $\nu : N_0 \to U$ are the state, output, and input signals, respectively.

We write $\zeta_{xv}(k)$ to denote the point reached at time $k$ under the input signal $\nu$ from initial condition $x = \zeta_{xv}(0)$. Similarly, we denote by $\zeta_{xv}(k)$ the output corresponding to state $\zeta_{xv}(k)$, i.e., $\zeta_{xv}(k) = h(\zeta_{xv}(k))$. In the above definition, we implicitly assumed that $X$ is positively invariant.]

Now, we introduce the notion of incremental input-to-state stability (\delta-ISS) leveraged later to show some of the main results of the paper.

**Definition VI.2.** [48] System $\Sigma = (X, S, U, f, Y, h)$ is called incrementally input-to-state stable (\delta-ISS) if there exist a $K\mathcal{L}$ function $\beta$ and $K_\infty$ function $\gamma$ such that for all $x, x' \in X$ and $\forall v, u' : N_0 \to U$, the following inequality holds for any $k \in \mathbb{N}$:

$$\|\zeta_{xv}(k) - \zeta_{x'v'}(k)\| \leq \beta(\|x - x'\|, k) + \gamma(\|v - u'\|, k).$$

**Example VI.3.** As an example, for a linear control system:

$$\begin{align*}
(\xi(k + 1) &= A\xi(k) + Bu(k), \\
h(k) &= C\xi(k),
\end{align*}$$

where all eigenvalues of $A$ are inside the unit circle, the functions $\beta$ and $\gamma$ can be chosen as:

$$\begin{align*}
\beta(r, k) &= \|A^r\| r, \\
\gamma(r, k) &= \|B\| \sum_{n=0}^{\infty} \|A^n\| r.
\end{align*}$$

In general, it is difficult to check inequality (15) directly for nonlinear systems. Fortunately, \delta-ISS can be characterized using Lyapunov functions.

**Definition VI.4.** [48] Consider a control system $\Sigma$ and a continuous function $V : X \times X \to \mathbb{R}_+^\infty$. Function $V$ is called a $\delta$-ISS Lyapunov function for $\Sigma$ if there exist $K_\infty$ functions $\alpha_1, \alpha_2, \rho$ and $K$ function $\sigma$ such that:

(i) for any $x, x' \in X$

$$\alpha_1(\|x - x'\|) \leq V(x, x') \leq \alpha_2(\|x - x'\|);$$

(ii) for any $x, x' \in X$ and $u, u' \in U$

$$V(f(x, u), f(x', u')) - V(x, x') \leq -\rho(\|V(x, x')\|)\sigma(\|u - u'\|);$$

The following result characterizes \delta-ISS in terms of existence of \delta-ISS Lyapunov functions.

**Theorem VI.5.** [48] Consider a control system $\Sigma$.

- $\Sigma$ is \delta-ISS if it admits a $\delta$-ISS Lyapunov function.
- If $U$ is compact and convex and $X$ is compact, then the existence of a $\delta$-ISS Lyapunov function is equivalent to $\delta$-ISS.

The next technical lemma will be used later to show some of the main results of this section.

1Set $X$ is called positively invariant under (14) if $\xi_{xv}(k) \in X$ for any $k \in \mathbb{N}$, any $x \in X$ and any $\nu : N_0 \to U$. 
Lemma VI.6. Consider a control system $\Sigma$. Suppose $V$ is a $\delta$-ISS Lyapunov function for $\Sigma$. Then there exist $\kappa, \lambda \in \mathbb{K}_\infty$, where $\kappa(s) < s$ for any $s \in \mathbb{R}^+$, such that
\[
V(f(x, u), f(x', u')) \leq \max\{\kappa(V(x, x')), \lambda(||u - u'||)\},
\]for any $x, x' \in \mathbb{X}$ and any $u, u' \in \mathcal{U}$.

The proof is similar to that of Theorem 1 in [49] and is omitted here due to lack of space.

In order to provide the main results of this section, we first describe control systems in Definition VI.1 as metric systems as in Definition II.1. More precisely, given a control system $\Sigma = (\mathcal{X}, \mathcal{S}, \mathcal{U}, f, \gamma, h)$, we define an associated metric system
\[
S(\Sigma) = (X, X_0, X_S, U, \gamma, Y, H),
\]where $X = \mathbb{X}$, $X_0 = \mathbb{X}$, $X_S = \mathcal{S}$, $U = \mathcal{U}$, $Y = \gamma$, $H = h$, and $x \overset{u}{\rightarrow} x'$ if and only if $x' = f(x, u)$. We assume that the output set $Y$ is equipped with the infinity norm: $d(y_1, y_2) = ||y_1 - y_2||_\infty$. We have a similar assumption for the state set $X$.

Now, we introduce a symbolic system for the control system $\Sigma = (\mathcal{X}, \mathcal{S}, \mathcal{U}, f, \gamma, h)$. To do so, from now on we assume that sets $\mathcal{X}$, $\mathcal{S}$ and $\mathcal{U}$ are of the form of finite union of boxes. Consider a concrete control system $\Sigma$ and a tuple $q = (\eta, \mu, \theta)$ of parameters, where $0 < \eta < \min \{\text{span}(\mathcal{S}), \text{span}(\mathcal{X} \setminus \mathcal{S})\}$ is the state set quantization, $0 < \mu < \text{span}(\mathcal{U})$ is the input set quantization, and $\theta$ is a design parameter. Now let us introduce the symbolic system
\[
S_q(\Sigma) = (X_q, X_0, X_S, U, \gamma, Y, H_q),
\]where $X_q = X \times [\mathcal{S}]^q$, $X_S = [\mathcal{S}]^q$, $U = \mathcal{U}$, $Y_q = \{h(x_q) | x_q \in X_q\}$, $H_q(x_q) = h(x_q)$, and $x_q \overset{u_q}{\rightarrow} x'_q$ if and only if $||x'_q - f(x_q, u_q)|| \leq \eta$.

We can now state the first main result of this section showing that, under some condition over the quantization parameters $\eta$ and $\mu$, $S_q(\Sigma)$ and $S(\Sigma)$ are related under an approximate initial-state opacity preserving simulation relation.

**Theorem VI.7.** Let $\Sigma = (\mathcal{X}, \mathcal{S}, \mathcal{U}, f, \gamma, h)$ be a $\delta$-ISS control system. For any desired precision $\varepsilon > 0$, and any tuple $q = (\eta, \mu, 0)$ of parameters satisfying
\[
\beta(\alpha^{-1}(\varepsilon), 1) + \gamma(\mu) + \eta \leq \alpha^{-1}(\varepsilon), \tag{21}
\]we have $S(\Sigma) \preceq_S S_q(\Sigma) \preceq_S S(\Sigma)$.

**Proof.** We start by proving $S(\Sigma) \preceq_S S_q(\Sigma)$.

Consider the relation $R \subset X \times X_q$ defined by $(x, x_q) \in R$ if and only if $||x - x_q|| \leq \alpha^{-1}(\varepsilon)$. Since $\eta \leq \text{span}(\mathcal{S})$, $X_S \subset \bigcup_{p \in [\mathcal{S}]^q} \mathcal{B}_p(\eta)$, and by (21), $\forall x \in X_q$, $\exists x \in X_S$ such that:
\[
||x - x_q|| \leq \eta \leq \alpha^{-1}(\varepsilon). \tag{22}
\]

Hence, $(x, x_q) \in R$ and condition 1)-a) in Definition VI.1 is satisfied. For every $x_q \in X_q \setminus X_S$, by choosing $x = x_q$ which is also inside set $X \setminus X_S$, one gets $(x, x_q) \in R$ and, hence, condition 1)-b) in Definition VI.1 holds as well. Now consider any $(x, x_q) \in R$. Condition 2) in Definition VI.1 is satisfied by the definition of $R$ and the Lipschitz assumption:
\[
||H(x) - H_q(x_q)|| = ||h(x) - h(x_q)|| \leq \alpha(||x - x_q||) \leq \varepsilon.
\]

Let us now show that condition 3) in Definition VI.1 holds.

Consider any $u \in \mathcal{U}$. Choose an input $u_q \in U_q$ satisfying:
\[
||u - u_q|| \leq \mu. \tag{23}
\]

Note that the existence of such $u_q$ is guaranteed by the inequality $\mu \leq \text{span}(\mathcal{U})$ which guarantees that $\mathcal{U} \subset \bigcup_{p \in \mathcal{S}} \mathcal{B}_p(\eta)$.

Consider the unique transition $x \overset{u}{\rightarrow} x' = f(x, u)$ in $S(\Sigma)$. It follows from the $\delta$-ISS assumption on $\Sigma$ and (23) that the distance between $x'$ and $f(x, u_q)$ is bounded as:
\[
||x' - f(x, u_q)|| \leq \beta(||x - x_q||, 1) + \gamma(||u - u_q||) \leq \beta(\alpha^{-1}(\varepsilon), 1) + \gamma(\mu). \tag{24}
\]

Since $X \subset \bigcup_{p \in [\mathcal{S}]^q} \mathcal{B}_p(\eta)$, there exists $x_q' \in X_q$ such that:
\[
||x'(x, u_q) - x_q'|| \leq \eta, \tag{25}
\]

which, by the definition of $S_q(\Sigma)$, implies the existence of $x_q \overset{u_q}{\rightarrow} x_q'$ in $S_q(\Sigma)$. Using the inequalities (21), (24), and triangle inequality, we obtain:
\[
||x' - x_q'|| \leq ||x' - f(x, u_q)|| + ||f(x, u_q) - x_q'|| \leq \beta(\alpha^{-1}(\varepsilon), 1) + \gamma(\mu) + \eta \leq \alpha^{-1}(\varepsilon).
\]

Therefore, we conclude $(x', x_q') \in R$ and condition 3)-a) in Definition VI.1 holds. Let us now show that condition 3)-b) in Definition VI.1 also holds.

Now consider any $(x, x_q) \in R$ and any $u_q \in U_q$. Choose the input $u = u_q$ and consider the unique $x' = f(x, u)$ in $S(\Sigma)$.

Using $\delta$-ISS assumption for $\Sigma$, we bound the distance between $x'$ and $f(x_q, u_q)$ as:
\[
||x' - f(x, u_q)|| \leq \beta(||x - x_q||, 1) \leq \beta(\alpha^{-1}(\varepsilon), 1). \tag{26}
\]

Using the definition of $S_q(\Sigma)$, the inequalities (21), (26), and the triangle inequality, we obtain:
\[
||x' - x_q'|| \leq ||x' - f(x_q, u_q)|| + ||f(x_q, u_q) - x_q'|| \leq \beta(\alpha^{-1}(\varepsilon), 1) + \eta \leq \alpha^{-1}(\varepsilon)
\]

Therefore, we conclude $(x', x_q') \in R$ and condition 3)-b) in Definition VI.1 holds.

In a similar way, one can prove that $S_q(\Sigma) \preceq_S S(\Sigma)$.

**Remark VI.8.** Note that there always exist quantization parameters $q$ such that inequality (21) holds as long as $\beta(\alpha^{-1}(\varepsilon), 1) < \alpha^{-1}(\varepsilon)$. By assuming that the discrete-time control system $\Sigma$ is a sampled-data version of an original continuous-time one with the sampling time $\tau$, one can ensure the latter inequality by choosing the sampling time large enough given that $\beta(\tau, 1) = \beta(\tau, r) < r$ for some $KL$ function $\beta$ establishing the incremental stability of the original continuous-time system. For example, for the function in (17), one has $\beta(\tau, 1) = ||A||\tau = ||e^{\tau A}||\tau$, where $A$ is the state matrix of the original continuous-time linear control system.

The following example illustrates how to use Theorem VI.7 to verify approximate opacity for an infinite system based on its finite abstraction.
Example VI.9. Let us consider the following simple system
\[\Sigma: \begin{cases} 
\xi(k+1) = 0.1\xi(k) + v(k), \\
\zeta(k) = \sin(2.5\pi\xi(k)) + 1,
\end{cases}\]  
where \(\mathbb{X} = [0, 1.6], S = [0, 0.1]\), and \(U = \{0.001\}\). This system is clearly \(\delta\)-ISS and according to Equation (17), we have \(\beta(r, k) = 0.1^k r\) and \(\gamma(r) = \sum_{m=0}^{\infty} 0.1^m r\). Also, function \(h\) satisfies the Lipschitz condition with \(\alpha(r) = 2.5\pi r\). By Equation (21), the parameters \(q = (\eta, \mu, 0)\) and the abstract precision \(\varepsilon\) should satisfy \(0.04\varepsilon + 10\frac{\mu}{\eta} + \eta \leq 0.4\varepsilon\). Let us consider desired abstract precision \(\varepsilon = 0.09\) and quantization parameters \(q = (\eta, \mu, 0) = (0.1, 0.001, 0)\) satisfying the inequality. Then we obtain symbolic system \(S_q(\Sigma)\) shown in Figure 4, and by Theorem VI.7, we have \(S(\Sigma) \leq_{0.9} S_q(\Sigma) \leq_{0.9} S(\Sigma)\). Essentially, we discretize the state space of \([0, 1.6]\) into 16 discrete states based on parameter \(\eta\). One can easily check that \(S_q(\Sigma)\) is 0-approximate initial-state opaque since for any run from secret initial state 0, there exists a run from non-secret state 8 such that their outputs are exactly the same. Therefore, by Theorem V.2, we can conclude that \(\Sigma\) is 1.8-approximate initial-state opaque.

The next theorem provides similar results as in Theorem VI.7 but by leveraging \(\delta\)-ISS Lyapunov functions. To show the next result, we will make the following supplementary assumption on the \(\delta\)-ISS Lyapunov functions: there exists a function \(\gamma \in \mathcal{K}_\infty\) such that
\[\forall x, x', x'' \in \mathbb{X}, \quad V(x, x') - V(x', x'') \leq \gamma(||x - x'||).\]  
Inequality (28) is not restrictive at all provided we are interested in the dynamics of the control system on a compact subset of the state set \(\mathbb{X}\); see the discussion in [34].

Theorem VI.10. Let \(\Sigma = (\mathbb{X}, S, \mathbb{U}, f, Y, h)\) admit a \(\delta\)-ISS Lyapunov function \(V\) satisfying (28). For any desired precision \(\varepsilon > 0\), and any tuple \(q = (\eta, \mu, 0)\) satisfying
\[\alpha_2(\eta) \leq \alpha_1(\alpha^{-1}(\varepsilon)), \quad \max\{\kappa(\alpha_1(\alpha^{-1}(\varepsilon))), \lambda(\mu)\} + \gamma(\eta) \leq \alpha_1(\alpha^{-1}(\varepsilon))\]  
we have \(S(\Sigma) \leq_{\varepsilon}^\delta S_q(\Sigma) \leq_{\varepsilon}^\delta S(\Sigma)\).

Proof. We start by proving \(S(\Sigma) \leq_{\varepsilon}^\delta S_q(\Sigma)\). Consider the relation \(R \subseteq X \times X_q\) defined by \((x, x_q) \in R\) if and only if \(V(x, x_q) \leq \alpha_1(\alpha^{-1}(\varepsilon))\). Since \(\eta \leq \text{span}(\mathbb{S})\) and \(X_S \subseteq \bigcup_{p \in [\mathbb{S}]} B_\eta(p)\), for every \(x \in X_S\) there always exists \(x_q \in X_{\alpha\eta}\) such that \(||x - x_q|| \leq \eta\). Then
\[V(x, x_q) \leq \alpha_2(||x - x_q||) \leq \alpha_2(\eta) \leq \alpha_1(\alpha^{-1}(\varepsilon))\]
because of (29) and \(\alpha_2\) being a \(\mathcal{K}_\infty\) function. Hence, \((x, x_q) \in R\) and condition 1-a) in Definition V.1 is satisfied. For every \(x_q \in X_S \setminus X_{\alpha\eta}\), by choosing \(x = x_q\) which is also inside set \(X \setminus X_S\), one gets trivially \((x, x_q) \in R\) and, hence, condition 1-b) in Definition V.1 holds as well. Now consider any \((x, x_q) \in R\). Condition 2) in Definition V.1 is satisfied by the definition of \(R\) and the Lipschitz assumption on map \(h\) as in Definition VI.1:
\[\|H(x) - H_q(x_q)\| = \|h(x) - h(x_q)\| \leq \alpha(||x - x_q||)\]
\[\leq \alpha_1^(-1)(V(x, x_q)) \leq \varepsilon.\]

Let us now show that condition 3) in Definition V.1 holds.
Consider any \(u \in U\). Choose an input \(u_q \in U_q\) satisfying:
\[\|u - u_q\| \leq \mu.\]

Note that the existence of such \(u_q\) is guaranteed by the inequality \(\mu \leq \text{span}(\mathbb{U})\) which guarantees that \(U \subseteq \bigcup_{p \in [\mathbb{S}]} B_\eta(p)\).

Consider the unique transition \(x \xrightarrow{u} x' = f(x, u)\) in \(S(\Sigma)\). Given \(\delta\)-ISS Lyapunov function \(V\) for \(\Sigma\), inequality (18), and (31), one obtains:
\[V(x', f(x_q, u)) \leq \max\{\kappa(V(x, x_q)), \lambda(||u - u_q||)\}\]
\[\leq \max\{\kappa(\alpha_1(\alpha^{-1}(\varepsilon))), \lambda(\mu)\}\]

Since \(X \subseteq \bigcup_{p \in [\mathbb{S}]} B_\eta(p)\), there exists \(x_q' \in X_S\) such that:
\[\|f(x_q, u) - x_q'\| \leq \eta,\]

which, by the definition of \(S_q(\Sigma)\), implies the existence of \(x_q \xrightarrow{u_q} x_q'\) in \(S_q(\Sigma)\). Using the inequalities (28), (30), and (33), we obtain:
\[V(x', x_q') \leq V(x', f(x_q, u)) + \gamma(||f(x_q, u) - x_q'||)\]
\[\leq \max\{\kappa(\alpha_1(\alpha^{-1}(\varepsilon))), \lambda(\mu)\} + \gamma(\eta) \]
\[\leq \alpha_1(\alpha^{-1}(\varepsilon)).\]

Therefore, we conclude \((x', x_q') \in R\) and condition 3-a) in Definition V.1 holds. Let us now show that condition 3-b) in Definition V.1 also holds.

Consider any \((x, x_q) \in R\). Consider any \(u_q \in U_q\). Choose the input \(u = u_q\) and consider the unique \(x' = f(x, u)\) in \(S(\Sigma)\). Given \(\delta\)-ISS Lyapunov function \(V\) for \(\Sigma\) and inequality (18), one gets:
\[V(x', f(x_q, u)) \leq \kappa(V(x, x_q)) \leq \kappa(\alpha_1(\alpha^{-1}(\varepsilon))).\]

Using the definition of \(S_q(\Sigma)\), the inequalities (28), (30), and (33), we obtain:
\[V(x', x_q') \leq V(x', f(x_q, u)) + \gamma(||f(x_q, u) - x_q'||)\]
\[\leq \kappa(\alpha_1(\alpha^{-1}(\varepsilon))) + \gamma(\eta) \leq \alpha_1(\alpha^{-1}(\varepsilon)).\]

Therefore, we conclude that \((x', x_q') \in R\) and condition 3-b) in Definition V.1 holds.

In a similar way, one can prove that \(S_q(\Sigma) \leq_{\varepsilon}^\delta S(\Sigma)\).
Remark VI.11. One can readily verify that there always exists a choice of quantization parameter \( q = (\eta, \mu) \) such that inequalities (29) and (30) hold simultaneously. Although the result in Theorem VI.10 seems more general than that of Theorem VI.7 in terms of the existence of quantization parameter \( q \), the symbolic model \( S_q(\Sigma) \), computed by using the quantization parameters \( q \) provided in Theorem VI.7 whenever existing, is likely to have fewer states than the model computed by using the quantization parameters provided in Theorem VI.10 due to the conservative nature of \( \delta \)-ISS Lyapunov functions.

The next theorems illustrate the other main results of this section showing that, under similar conditions over the quantization parameters \( \eta \) and \( \mu \), \( S_q(\Sigma) \) and \( S(\Sigma) \) are related under an approximate current-state opacity preserving simulation relation.

Theorem VI.12. Let \( \Sigma = (X, S, U, f, Y, h) \) be a \( \delta \)-ISS control system. For any desired precision \( \varepsilon > 0 \), and any tuple \( q = (\eta, \mu, \theta) \) of parameters satisfying
\[
\beta (\alpha^{-1}(\varepsilon), 1) + \gamma (\mu) + \eta \leq \alpha^{-1}(\varepsilon), \\
\alpha^{-1}(\varepsilon) \leq \theta,
\]
we have \( S(\Sigma) \preceq_{S} S_q(\Sigma) \).

Proof. Consider the relation \( R \subseteq X \times X_q \) defined by \( (x, x'_q) \in R \) if and only if \( \|x - x'_q\| \leq \alpha^{-1}(\varepsilon) \). Note that conditions 1), 2), 3)-a) and 3)-c) of \( \varepsilon \)-CURSOP simulation relation in Definition V.6 are similar to that of \( \varepsilon \)-InitSOP simulation relation, therefore the proof of them are similar to that in Theorem VI.7 and is omitted here. Here, we show that conditions 3)-b) and 3)-d) in Definition V.6 hold.

Let us consider an arbitrary transition \( x \xrightarrow{u} x' = f(x, u) \) with \( x' \in X_S \) in \( S(\Sigma) \). Similar to the proof of condition 3)-a), we can show the existence of a transition \( x'_q \xrightarrow{u_q} x'_q \) in \( S_q(\Sigma) \) where \( (x'_q, x'_{q'}) \in R \) holds, where the input \( u_q \in U_q \) satisfies: \( \|u - u_q\| \leq \mu \). By the construction of the secret set in the symbolic system, one has \( X_qS = [0^\theta] \) with \( \theta \geq \alpha^{-1}(\varepsilon) \) and \( 0 < \eta \leq \min \{\text{span}(S), \text{span}(X \setminus S)\} \). Therefore, since \( (x'_q, x'_{q'}) \in R \) which implies \( \|x' - x'_{q'}\| \leq \alpha^{-1}(\varepsilon) \), we obtain that \( x'_{q'} \in X_qS \). Thus, we conclude that condition 3)-b) in Definition V.6 holds. In a similar way, we can show that condition 3)-d) in Definition V.6 holds as well which completes the proof.

Theorem VI.13. Let \( \Sigma = (X, S, U, f, Y, h) \) admits a \( \delta \)-ISS Lyapunov function \( V \) satisfying (28). For any desired precision \( \varepsilon > 0 \), and any tuple \( q = (\eta, \mu, \theta) \) satisfying
\[
\alpha_2(\eta) \leq \alpha_1(\alpha^{-1}(\varepsilon)), \\
\max \{\kappa(\alpha_1(\alpha^{-1}(\varepsilon))), \lambda(\mu)\} + \gamma (\eta) \leq \alpha_1(\alpha^{-1}(\varepsilon)), \\
\alpha^{-1}(\varepsilon) \leq \theta,
\]
we have \( S(\Sigma) \preceq_{S} S_q(\Sigma) \).

Proof. The proof is similar to that of Theorem VI.10 and VI.12 and is omitted here due to lack of space.

Since we show \( S(\Sigma) \preceq_{S} S_q(\Sigma) \) and \( S(\Sigma) \preceq_{S} S_q(\Sigma) \) under the same relation in Theorems VI.7 and VI.12 (resp. Theorems VI.10 and VI.13), by the definition of approximate infinite-state opacity preserving simulation relation, we consequently get the following results.

Theorem VI.14. Let \( \Sigma = (X, S, U, f, Y, h) \) be a \( \delta \)-ISS control system. For any desired precision \( \varepsilon > 0 \), and any tuple \( q = (\eta, \mu, \theta) \) of parameters satisfying
\[
\beta (\alpha^{-1}(\varepsilon), 1) + \gamma (\mu) + \eta \leq \alpha^{-1}(\varepsilon), \\
\alpha^{-1}(\varepsilon) \leq \theta,
\]
we have \( S(\Sigma) \preceq_{S} S_q(\Sigma) \).

Theorem VI.15. Let \( \Sigma = (X, S, U, f, Y, h) \) admits a \( \delta \)-ISS Lyapunov function \( V \) satisfying (28). For any desired precision \( \varepsilon > 0 \), and any tuple \( q = (\eta, \mu, \theta) \) satisfying
\[
\alpha_2(\eta) \leq \alpha_1(\alpha^{-1}(\varepsilon)), \\
\max \{\kappa(\alpha_1(\alpha^{-1}(\varepsilon))), \lambda(\mu)\} + \gamma (\eta) \leq \alpha_1(\alpha^{-1}(\varepsilon)), \\
\alpha^{-1}(\varepsilon) \leq \theta,
\]
we have \( S(\Sigma) \preceq_{S} S_q(\Sigma) \).

VII. CONCLUSION

In this paper, we extended the concept of opacity to metric systems by proposing the notion of approximate opacity. Verification algorithms and approximate relations that preserve approximate opacity were also provided. We also discussed how to construct finite abstractions that approximately simulate a class of control systems in terms of opacity preservation. Our result bridges the gap between the opacity analysis of finite discrete systems and continuous control systems.

Among the many possible directions for future work that will be built based on the proposed framework, we mention several directions of immediate interest. One direction is to extend our framework to the stochastic setting for almost opacity [14], [21]–[23]. Also, we are interested in constructing approximate opacity preserving symbolic models for more classes of systems. Finally, we plan to extend approximate opacity preserving simulation relation to approximate opacity preserving alternating simulation relation [45] and solve the problem of controller synthesis enforcing approximate opacity [46], [50]–[54].

APPENDIX

A. Proofs not contained in main body

Proof of Proposition IV.2

Proof. It is straightforward to show (i). Hereafter, we prove (ii) by induction on the length of input sequence.

When \( n = 0 \), i.e., there is no input sequence, we have that \( (x_0, q_0) \in X_{f0} \). By the definition of \( X_{f0} \), we know that \( q_0 = \{x'_0 \in X : d(H(x_0), H(x'_0)) \leq \delta\} \) which implies (ii) immediately.

To proceed the induction, we assume that (ii) holds when \( n = k \). Now, we need to show that (ii) also holds when \( n = k + 1 \). Consider arbitrary pair \( (x_0, q_0) \in X_{f0} \) and finite run \( (x_0, q_0) \xrightarrow{u_1} (x_1, q_1) \xrightarrow{u_2} \cdots \xrightarrow{u_n} (x_n, q_n) \xrightarrow{u_{n+1}} (x_{n+1}, q_{n+1}) \).
Then, we have
\[ q_{n+1} = \bigcup_{u \in U} \text{Pre}_2(q_n) \cap \{ x \in X : d(H(x_{n+1}), H(x)) \leq \delta \} \]
\[ = \bigcup_{x \in X : \exists x' \in q_n, u_{n+1} \in U \text{ s.t. } (x, u_{n+1}, x') \in \rightarrow} \]
\[ \cap \{ x \in X : d(H(x_{n+1}), H(x)) \leq \delta \} \]
\[ = \bigcup_{x \in X : \exists x' \in q_n, u_{n+1} \in U \text{ s.t. } (x, u_{n+1}, x') \in \rightarrow} \]
\[ \bigwedge d(H(x_{n+1}), H(x)) \leq \delta \]

By the induction hypothesis, we know that
\[ q_n = \left\{ x'_0 \in X : \exists x'_0, u'_0 \in [0,1,...,n] \text{ such that } d(H(x'_0), H(x'_0)) \leq \delta \right\}. \]

Therefore, by combining the above two equations, one gets
\[ q_{n+1} = \left\{ x \in X : \exists x'_0, u'_0 \in [0,1,...,n] \text{ such that } d(H(x'_0), H(x'_0)) \leq \delta \right\}. \]

By Proposition IV.6, we know that
\[ q_n \cap q'_n = \left\{ x'_0 \in X : \exists x'_0, u'_0 \in [0,1,...,n] \text{ such that } d(H(x'_0), H(x'_0)) \leq \delta \right\}. \]

Since \( q_n \cap q'_n \subseteq X_S \), we know that there does not exist \( x'_0 \in X_S \) and \( x'_0 \in [0,1,...,n] \) such that \( x'_0 \in X \setminus X_S \) and \( d(H(x'_0), H(x'_0)) \leq \delta \). Therefore, the system is not \( \delta \)-approximate infinite-step opaque.

By Proposition IV.6 and IV.7, we have \( x_0 \xrightarrow{u_0} \cdots \xrightarrow{u_{n-1}} x_{n-1} \xrightarrow{u_n} x_n \) and \( x_0 \xrightarrow{u_0} \cdots \xrightarrow{u_{n-1}} x_{n-1} \xrightarrow{u_n} x_n \) such that \( x'_n \in X \setminus X_S \) and \( d(H(x'_n), H(x'_n)) \leq \delta \). Let us consider the following sequence of transitions in \( S_C \)
\[ (x_0, q_0) \xrightarrow{u_0} (x_1, q_1) \xrightarrow{u_2} \cdots \xrightarrow{u_n} (x_n, q_n) \]

By Proposition IV.2, we know that
\[ q_n = \left\{ x'_0 \in X : \exists x'_0, u'_0 \in [0,1,...,n] \text{ such that } d(H(x'_0), H(x'_0)) \leq \delta \right\}. \]

By Equation (3), we have \( q_n \cap X_0 \subseteq X_S \). Therefore, there exists an initial state \( x'_0 \in X_0 \setminus X_S \) and a sequence \( x'_0 \xrightarrow{u_0} x'_1 \xrightarrow{u_2} \cdots x'_n \) such that \( d(H(x'_n), H(x'_n)) \leq \delta \). This is a contradiction, i.e., \( S \) has to be \( \delta \)-approximate initial-state opaque.
such that $x_k \in X \setminus X_S$ and
$$ \max_{i \in \{0, 1, \ldots, n\}} d(H(x_i), H(x'_0)) \leq \delta , $$
which is a contradiction, i.e., $S$ has to be $\delta$-approximate infinite-step opaque.

REFERENCES


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