Reachability Analysis for Solvable Dynamical Systems

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The reachability problem is one of the most important issues in the verification of hybrid systems. But unfortunately the reachable sets for most of hybrid systems are not computable. In the literature, only some special families of linear vector fields are proved with decidable reachability problem, let alone non-linear ones. In this paper, we investigate the reachability problem of non-linear vector fields by identifying three families of non-linear vector fields with solvability and prove their reachability problems are decidable. An *n*-dimension dynamical system is called *solvable* if its state variables can be partitioned into *m* groups such that the derivatives of the variables in the *i*th group are linear in themselves, but possibly non-linear in the variables from the 1st to i-1th groups. The three families of non-linear solvable vector fields under consideration are: the matrices corresponding to the linear parts of any vector field in the first family are nilpotent; the matrices corresponding to the linear parts of any vector in the second family are only with real eigenvalues; the matrices corresponding to the linear parts of any vector with pure imaginary eigenvalues. The experimental results indicate the efficiency of our approach.

Index Terms-Hybrid systems, Reachability analysis, Solvable systems, Tarski's algebra

I. INTRODUCTION

Hybrid systems (HSs) integrate computation with physical processes: embedded computers and networks monitor and control physical processes and feedback loops continuously influence computations, which are known as Cyber-Physical Systems (CPSs) nowadays. Applications of CPS span over many safety-critical domains, e.g., communication, healthcare, manufacturing, aerospace, transportation, etc. To guarantee the correctness of these systems is vital so that we can bet our lives on them, and challenging [40]. Therefore, formal methods has been widely used in the verification of HSs. The reachability problem of HSs is to verify that unsafe states are not reachable from the set of the initial states for a given HS, which is one of most important issues in the verification of HSs.

As HSs consist of intangibly interaction between continuous evolutions and discrete transitions, the reachability problem of most of HSs is undecidable [21], except for some simple cases, either their vector fields, *i.e.*, their continuous evolution parts, are quite simple such as timed automata [4] and multi-rate automata [3], or there are very restrictive constraints on their discrete transitions like o-minimal HSs [26].

In [27], Lafferriere et al. investigated vector fields of the form

$$\boldsymbol{\xi} = \boldsymbol{A}\boldsymbol{\xi} + \mathbf{u},\tag{1}$$

where $\xi(t) \in \mathbb{R}^n$ is the state of the system at time $t, A \in \mathbb{R}^{n \times n}$ is the system matrix, and $\mathbf{u} : \mathbb{R} \to \mathbb{R}^n$ is a piecewise continuous function which is called the *input*. They obtained the decidability of the reachability problems of the following three families of vector fields:

- 1) A is *nilpotent*, *i.e.* $A^n = 0$, and each component of **u** is a polynomial;
- 2) A is *diagonalizable* with rational eigenvalues, and each component of **u** is of the form $\sum_{i=1}^{m} c_i e^{\lambda_i t}$, where λ_i s are rationals and c_i s are subject to semi-algebraic constraints;
- A is *diagonalizable* with purely imaginary eigenvalues, whose imaginary parts are rationals, and each component of **u** of the form ∑_{i=1}^m c_i sin(λ_it) + d_i cos(λ_it), where λ_is are rationals and c_is and d_is are subject to semi-algebraic constraints.

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The above results are achieved by reducing the problems into Tarski's algebra [39].

In [5], Anai and Weispfenning presented a systematic approach on how to reduce the reachability problem and control parameter set problem of parametric inhomogeneous linear differential systems, with the form

$$\dot{\boldsymbol{\xi}} = \boldsymbol{A}\boldsymbol{\xi} + \mathbf{u}(t, \mathbf{r}),^{1} \tag{2}$$

where $A \in \mathbb{R}^{n \times n}$ is an $n \times n$ matrix, $\mathbf{r} = (r_1, \dots, r_k)$ is a vector of parameters, to the transcendental implicitization problem of a fundamental system of solutions of $\dot{\xi} = A\xi$ by quantifier elimination. They further proved (Corollary 2 of [5]) that exact semi-algebraic implicitization is possible for a fundamental system of solutions of $\dot{\xi} = A\xi$ if and only if one of the following cases holds:

- 1') All eigenvalues of A are zero, i.e., A is nilpotent.
- All eigenvalues λ₁,..., λ_n of A are non-zero, pairwise distinct reals, and dim_Q(span(λ₁,..., λ_n)) ≤ 1.
- All eigenvalues λ₁,..., λ_n of A are purely imaginary, say of the form λ_i = μ_ii with non-zero, pairwise distinct reals μ_is, and dim_Q(span(μ₁,..., μ_n)) ≤ 1.

Obviously, Anai and Weispfenning's work extended Lafferriere *et al.*'s further, and particularly proved the largest families of linear vector fields whose exact reachable set computations are computable by reduction to Tarski's algebra.

In [16], [17], we extended the decidability results of reachability problems of linear vector fields due to Lafferriere *et al.* [27] and Anai and Weispfenning [5].

- In [16], we generalized the above cases 2) and 2') to
 - A is diagonalizable with real eigenvalues, and each component of **u** is of the form $\sum_{i=1}^{m} c_i e^{\lambda_i t}$, where λ_i s are reals and c_i s are subject to semi-algebraic constraints.

Note that compared with the case 2') in [5], we dropped the constraint dim_Q(span($\lambda_1, \dots, \lambda_n$)) ≤ 1 , which restrict the eigenvalues to be linearly dependent over \mathbb{Q} . Such extension is substantial, since the new family is strictly more expressive, whose reachability problem cannot be essentially reduced to Tarski's algebra any more as in [27], [5]. To obtain the decid-

¹This form can be generalized to $\dot{\xi} = A(t)\xi + \mathbf{u}(t,\mathbf{r})$.

ability, we have to resort to the decidability of the extension of Tarski's algebra with functions of the form

$$f(t, \mathbf{x}) = \sum_{i=0}^{m} f_i(t, \mathbf{x}) e^{\lambda_i t},$$
(3)

where $m \in \mathbb{N}$, $f_i(t, \mathbf{x}) \in \mathbb{R}[t, \mathbf{x}]$, $\lambda_i \in \mathbb{R}$, $i = 0, 1, \dots, m$, and *e* is an irrational and transcendental number approximately equal to 2.718281828459. We denote the extension by \mathscr{T}_e .

- In [17], we generalized the above cases 3) and 3') to
 - *A* is *diagonalizable* with purely imaginary eigenvalues, whose imaginary parts are reals, and each component of **u** is of the form $\sum_{i=1}^{m} c_i \sin(\lambda_i t) + d_i \cos(\lambda_i t)$, where λ_i s are reals and c_i s and d_i s are subject to semi-algebraic constraints.

This is achieved still by reducing the decidability to Tarski's algebra [39] using the density results in number theory [20], rather either by direct replacement like [27] or by reduction to the transcendental implicitization problem like [5]. Note that compared to the case 3') in [5], we dropped the constraint dim_{\mathbb{Q}}(span(μ_1, \dots, μ_n)) ≤ 1 .

It is also worth noting that for linear vector fields, some other problems which are quite related to the reachability problem, have been investigated and proved to be decidable in the literature, such as the Polytope Escape Problem [32], Recurrent Reachability Problem [8], and the Skolem Problem [9]. But a main restriction on all of the results is that the unsafe set should only be linear and represented as a polyhedra, while in our results, the unsafe set can be non-linear and represented by a semi-algebraic set. For an effective verification method for the reachability problem of the former case, we refer to Yazarel and Pappas's work [43].

Tarski's algebra is the first-order theory of reals over the structure $\langle \mathbb{R}; +, -, \cdot, 0, 1 \rangle$, which is also called the elementary algebra and geometry. In [39], Tarski showed the decidability of Tarski's algebra. But whether the extension of Tarski's algebra with exponentiation over real closed fields is decidable (so-called "Tarski's conjecture") is still open. In [2], [30], Weispfenning *et al.* gave a partial solution to Tarski's algebra by allowing the decidability of the extension of Tarski's algebra by allowing terms of the form $f(t, \mathbf{x}, e^t)$, where $f(t, \mathbf{x}, y) \in \mathbb{R}[t, \mathbf{x}, y]$. In [41], Xu *et al.* considered how to generalize Weispfenning *et al.*'s approach by allowing functions of the form (3), but with the restriction that all the λ_i s are nonnegative integers. Obviously, \mathcal{T}_e is strictly more expressive than the ones considered in [2], [30], [41].

In the literature, there is very little decidability results on the reachability problems of non-linear vector fields. The first decidability results are given in [42] on the reachability problems for some specific solvable non-linear vector fields, which are proper subsets of the second family below we consider, by exploiting Weispfenning *et al.*'s result on Tarski's conjecture [2]. In this paper, we investigate this issue by identifying three families of solvable vector fields and proving their reachability problems are decidable by exploiting the techniques developed in our previous work [16], [17], which are the three largest non-linear vector fields with decidable reachability to the best of our knowledge.

The notion of *solvability* was first proposed in [35] for a class of polynomial programs, and was extended to dynamical and hybrid systems in [42]. Formally, a dynamical system

$$\dot{\boldsymbol{\xi}} = F(\boldsymbol{\xi}, \mathbf{u}(t))$$

is called *solvable system* (SS) if the variable vector $\xi = (\xi_1, \dots, \xi_n)$ can be classified into *m* groups $(m \le n)$

$$\zeta_1=(\xi_{11},\ldots,\xi_{1n_1}),\ldots,\zeta_m=(\xi_{m1},\ldots,\xi_{mn_m}),$$

and the dynamical system can be represented as the form:

$$\dot{\boldsymbol{\xi}} = \begin{bmatrix} \zeta_1 \\ \dot{\zeta}_2 \\ \vdots \\ \dot{\zeta}_m \end{bmatrix} = \begin{bmatrix} A_1 \zeta_1 + \mathbf{u}_1(t) \\ A_2 \zeta_2 + \mathbf{u}_2(t, \zeta_1) \\ \vdots \\ A_m \zeta_m + \mathbf{u}_m(t, \zeta_1, \dots, \zeta_{m-1}) \end{bmatrix}, \quad (4)$$

where $0 < n_1 < ... < n_m = n$ are integers, $m \in \mathbb{N}$, $A_1, ..., A_m$ are real matrices with corresponding dimensions, $\mathbf{u}_1, ..., \mathbf{u}_m$ are *polynomial-exponential-trigonometric functions* (PETFs, the definition will be given later). *E.g.*

$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{z} \end{bmatrix} = \begin{bmatrix} x + e^{-t} \\ 2y + x^2 - e^{-\sqrt{2}t} \\ \sqrt{3}z + xy + 2e^{-t} \end{bmatrix},$$
(5)

is a solvable system which is beyond the expression of the linear system. Obviously, all linear dynamical systems (LDSs) are also SSs.

Thus, the main contributions of this paper can be summarized as follows:

- 1") If A_1, \ldots, A_m in (4) are *nilpotent*, *i.e.* $A_1^{k_1} = 0, \ldots, A_m^{k_m} = 0$, for some $k_1, \ldots, k_m \in \mathbb{N}$, and each component of \mathbf{u}_i is a polynomial, then the reachability problem of (4) is decidable. This is achieved by reduction to Tarski's algebra similarly to [27].
- 2") If each A_i is *diagonalizable* with real eigenvalues, and each component of \mathbf{u}_i is of the form $\sum_{j=1}^{m_i} c_{ij}e^{\lambda_{ij}t}$, where λ_{ij} s are reals and c_{ij} s are subject to semi-algebraic constraints, then the reachability problem of (4) is decidable, where $i = 1, \ldots, m$. The technique adopted for this case is adapted from [16]. In [16], it is assumed that any expression of \mathcal{T}_e has no multiple real roots, we will drop such restriction in this paper.
- 3") If each A_i is *diagonalizable* with purely imaginary eigenvalues, whose imaginary parts are reals, and each component of \mathbf{u}_i of the form $\sum_{j=1}^{m_i} c_{ij} \sin(\lambda_{ij}t) + d_{ij} \cos(\lambda_{ij}t)$, where λ_{ij} s are reals and c_{ij} s and d_{ij} s are subject to semi-algebraic constraints, the reachability problem of (4) is decidable, where $i = 1, \dots, m$. The technique adopted for this case is essentially same as what we used in [17], but the reduction procedure is more complicated for the non-linear case.

Additionally, similar to [5], [17], we present an abstraction of general solvable dynamical systems of the form (4). That is,

• each A_i is a real matrix, and each component of \mathbf{u}_i is of the form $\sum_{k=0}^{r_i} p_{ik}(t) \exp^{\alpha_{ik}t} \cos(\beta_{ik}t + \gamma_{ik})$, where $i = 1, \ldots, m, r_i \in \mathbb{N}, \alpha_{ik}, \beta_{ik}, \gamma_{ik} \in \mathbb{R}$ and $p_{ik}(t) \in \mathbb{R}[t]$.

The basic idea of our approach is as follows: for each eigenvalue $\alpha \pm \beta i$ of A_i , we introduce two fresh variables *a* and *b*, and let $a = \sin\beta t$ and $b = \cos\beta t$. So, it derives a new constraint $a^2 + b^2 = 1$. Using such replacement, the reachable set of (4) can be essentially represented as the form

$$f(t, \mathbf{x}, \mathbf{a}, \mathbf{b}) = \sum_{i=0}^{m} \sum_{j=0}^{n_i} f_{ij}(t, \mathbf{x}, \mathbf{a}, \mathbf{b}) e^{\alpha_{ij}t}.$$

Clearly, constraints over such expressions together with all the derived constraints fall into the decidable theory \mathcal{T}_e .

We implement a prototypical tool of our approach, and some case studies are conducted. To demonstrate the efficiency of our approach, first, we compare our tool with CT1D [38], a generalized CAD implementation of Mathematica's Reduce command, which can cope with quantifier elimination of \mathcal{T}_e . For \mathcal{T}_e formulas only with strict inequalities, our tool outperforms CT1D, and for the rest cases, their efficiencies are nearly same. As other state-of-the-art tools for quantifier elimination, e.g., REDLOG [14], QEPCAD, and SyNRAC [24] cannot handle the decidability problems we considered in this paper in general, it is thus not comparable. Second, we also compare our tool with several well-known reachability computation tools based on approximation and numeric computation, e.g., HSolver [33], FLOW* [7], dReach [25], etc., although such comparisons are not fairly as they deal with different problems in general. After necessary preprocessing in order to make the comparison reasonable, our tool is more efficient.

The rest of this paper is organized as follows: In Section II, we first introduce some basic notions and theories, then explain the problem we consider. Section III gives a decision procedure to the reachability problem when all A_i s in (4) are nilpotent. Section IV is devoted to the case when all A_i s are only with real eigenvalues in (4). Section V is devoted to the case when all A_i s are only with purely imaginary eigenvalues in (4). In Section VI, we present an abstraction of general solvable non-linear differential systems. In Section VII, a prototypical implementation and experiments are reported. Section VIII concludes this paper and discusses future work.

II. PRELIMINARIES

In this section, we first introduce some basic notions and theories, then explain the problem we consider. We use \mathbf{x} to stand for a vector variable (x_1, \ldots, x_n) , $\mathbb{N}, \mathbb{Q}, \mathbb{R}, \mathbb{C}$ for natural, rational, real and complex numbers respectively, $\mathbb{R}[\mathbf{x}]$ for the polynomial ring in \mathbf{x} with coefficients in \mathbb{R} in what follows. We denote by $\Lambda(M)$ the set of all the eigenvalues of matrix M. For any $c \in \mathbb{C}$, denote by Im(c) the imaginary part of c.

A. Basic Notions

A term $f(t, \mathbf{x})$ is called *polynomial-exponential function* (PEF) w.r.t. t if it can be written in the form of (3).

A term $f(t, \mathbf{x})$ is called *trigonometric function* (TMF) w.r.t. t if it can be written with the following form:

$$f(t, \mathbf{x}) = \sum_{l=1}^{r} c_l(\mathbf{x}) cos(\boldsymbol{\mu}_l t) + d_l(\mathbf{x}) sin(\boldsymbol{\mu}_l t),$$
(6)

where $r \in \mathbb{N}, c_l(\mathbf{x}), d_l(\mathbf{x}) \in \mathbb{R}[\mathbf{x}]$ and $\mu_l \in \mathbb{R}, l = 1, \dots, r$. Denote by $\Gamma(f(t,\mathbf{x}))$ the set $\{\mu_1,\mu_2,\ldots,\mu_r\}$ in the sequel.

A term $f(t, \mathbf{x})$ is called a *polynomial-exponential-trigonometric function* (PETF) w.r.t. t, if it can be written with the following form:

$$f(t, \mathbf{x}) = \sum_{k=0}^{\prime} p_k(t, \mathbf{x}) e^{\alpha_k t} \cos(\beta_k t + \gamma_k), \tag{7}$$

where $r \in \mathbb{N}$, $\alpha_k, \beta_k, \gamma_k \in \mathbb{R}$ and $p_k(t, \mathbf{x}) \in \mathbb{R}[t, \mathbf{x}]$. Obviously, PEFs and TMFs are PETFs as $\sin(\alpha)$ can be seen as $\cos(\frac{\pi}{2} + \alpha)$.

A function vector is said to be PEF (TMF or PETF) if every component is a PEF (TMF or PETF).

A set $X \subset \mathbb{R}^n$ is said to be *semi-algebraic* if it is defined as

$$\mathbf{X} = \{ \mathbf{x} \in \mathbb{R}^n \mid p_1(\mathbf{x}) \triangleright 0, \cdots, p_j(\mathbf{x}) \triangleright 0 \},\$$

for some polynomials $p_1(\mathbf{x}), \dots, p_i(\mathbf{x}) \in \mathbb{R}[\mathbf{x}]$, where $\triangleright \in \{\geq, >\}$ and $j \in \mathbb{N}$. X is called *open semi-algebraic* if there is a ball $b_{\delta}(\mathbf{x})$ such that $b_{\delta}(\mathbf{x}) \subseteq \mathbf{X}$, where δ is the radius and \mathbf{x} is the center of the ball, for any $\mathbf{x} \in \mathbf{X}$.

B. Density Results in Number Theory

In this part, we introduce some theoretical results on density in number theory.

Definition 1 (Rational Linear Independent). Let a_1, \ldots, a_k are some real numbers. We say a_1, \ldots, a_k are rational linear independent if $\sum_{i=1}^{k} c_i a_i = 0 \text{ implies } \bigwedge_{i=1}^{k} c_i = 0, \text{ for all } c_1, \dots, c_k \in \mathbb{Q}.$

Definition 2 (Basis). Let $A \subset \mathbb{R}$ with $\#(A) < +\infty$ be a set of real numbers, where #(A) means the number of elements in A. A set $B \subseteq$ A is said be a basis of A, if the elements in B are rational linear independent and for any element $a \in A \setminus B$, where $A \setminus B$ denotes the set of all the elements in A but not in B, then the elements in $\{a\} \cup B$ are not rational linear independent.

Let $A = \{a_1, \dots, a_k\}$ be a set of real number, $B = \{b_1, \dots, b_j\} \subseteq A$ be a basis of A. It is easy to see that for any $a_i \in A$, there exists $c = (c_{i1}, \ldots, c_{ij}) \in \mathbb{Q}$ such that

$$a_i = c_{i1}b_1 + \ldots + c_{ij}b_j.$$
 (8)

For $1 \leq l \leq j$, let

$$d_l = \operatorname{lcm}(\operatorname{denom}(c_{1l}), \dots, \operatorname{denom}(c_{kl})), \tag{9}$$

where denom(c) is the denominator of rational number c and lcm means the least common multiple. Let $\overline{B} = \{\frac{b_1}{d_1}, \dots, \frac{b_j}{d_j}\}$ be a basis of A, then for any $a \in A$, a can be written as an integer linear combination of the elements in \overline{B} . We call such basis \overline{B} an *integer*basis of A.

The following Kronecker Theorem gives a nice density property of a rational or integer linear independent set [20].

Theorem 1 (Kronecker). The set $\{(\{\xi_1t\}_1,\ldots,\{\xi_kt\}_1) \mid t \in \mathbb{N}\}$ is dense in $[0,1]^k$, if $1,\xi_1,\ldots,\xi_k$ are integer linear independent, where $\{\xi\}_1 \in [0,1)$ is the decimal part of the real number ξ .

Corollary 1. The set $\{(\{\xi_1t\}_{2\pi},...,\{\xi_kt\}_{2\pi}) \mid t \ge 0\}$ is dense in $[0,2\pi]^k$, if ξ_1,\ldots,ξ_k are integer linear independent, where $\{\xi\}_{2\pi} \in$ $[0,2\pi)$ is the remainder of ξ by 2π .

Proof. Let $\xi'_i = \frac{\xi_i}{2\pi}$, for i = 1, ..., k. It is easy to see that we just need to prove that

$$\{(\{\xi_1't\}_1, \dots, \{\xi_k't\}_1) \mid t \ge 0\}$$
(10)

is dense in $[0,1]^k$. Since ξ_1, \ldots, ξ_k are integer linear independent, ξ'_1, \ldots, ξ'_k are also integer linear independent. Thus, it is easy to see that there exists $\xi_0 > 0$ such that $1, \xi_0 \xi'_1, \dots, \xi_0 \xi'_k$ are integer linear independent. By Theorem 1, it follows

$$\{(\{\xi_0\xi_1'n\}_1,\ldots,\{\xi_0\xi_k'n\}_1) \mid n \in \mathbb{N}\}$$
(11)

is dense in $[0,1]^k$. As $\xi_0 > 0$ implies $\{\xi_0 n \mid n \in \mathbb{N}\} \subset \{t \mid t \ge 0\}$, we have that the set in (11) is a subset of the set in (10). Thus, the set in (10) is dense in $[0, 1]^k$. \square

Theorem 2. Let a_1, \ldots, a_k be rational linear independent, and

$$S = \{(\sin(a_1t), \cos(a_1t), \dots, \sin(a_kt), \cos(a_kt)) \mid t \ge 0\},$$
(12)

$$\overline{S} = \{ (\alpha_1, \beta_1, \dots, \alpha_k, \beta_k) \in \mathbb{R}^{2k} \mid \bigwedge_{i=1}^{n} \alpha_i^2 + \beta_i^2 = 1 \},$$
(13)

then S is dense in \overline{S} .

Proof. a_1, \ldots, a_k are rational linear independent, then also integer linear independent. By Corollary 1, we have that

$$D_0 = \{ (\{a_1t\}_{2\pi}, \dots, \{a_kt\}_{2\pi}) \mid t \ge 0 \}$$

is dense in $D = [0, 2\pi]^k$. On the other hand, obviously, (\sin, \cos) : $D_0 \mapsto S$, and $(\sin, \cos) : D \mapsto \overline{S}$, and (\sin, \cos) is continuous, hence $(\sin, \cos)(D_0)$ is dense in $(\sin, \cos)(D)$, i.e., S is dense in \overline{S} .

Corollary 2. Let $f(\alpha_1, \beta_1, \dots, \alpha_k, \beta_k)$ be a polynomial in $\alpha_1, \beta_1, \ldots, \alpha_k, \beta_k$. a_1, \ldots, a_k are real numbers that are rational linear independent and S,\overline{S} defined as (12),(13), then f(S) is dense in $f(\overline{S})$.

Proof. By Theorem 2 we have S is dense in \overline{S} , and S, \overline{S} are both bounded sets. Since f is a polynomial, i.e. f is continuous, it is easy to see that f(S) is dense in $f(\overline{S})$.

C. Problem

Given an SS of the form (4) and an initial state $\xi(0) = \mathbf{x}$, the solution of this system at time $t \ge 0$ is denoted by $\xi(t) = \Phi(\mathbf{x}, t)$. Then the forward reachable set Post(X) of (4) from a given set X is defined as follows:

$$Post(\mathbf{X}) = \{ \mathbf{y} \in \mathbb{R}^n \mid \exists \mathbf{x} \exists t : \mathbf{x} \in \mathbf{X} \land t \ge 0 \land \Phi(\mathbf{x}, t) = \mathbf{y} \}.$$
(14)

The safety problem is: given an initial set X and an unsafe set Y, verify whether any unsafe state in Y is not reachable by some trajectory starting from X, *i.e.*, whether $Post(X) \cap Y = \emptyset$. Let

$$\mathscr{F}(\mathbf{X},\mathbf{Y}) = \exists \mathbf{x} \exists \mathbf{y} \exists t : \mathbf{x} \in \mathbf{X} \land \mathbf{y} \in \mathbf{Y} \land t \ge 0 \land \Phi(\mathbf{x},t) = \mathbf{y}.$$
(15)

That is, the safety problem is to verify whether the formula $\mathscr{F}(X, Y)$ is true or false. If it is false then the safety property holds, otherwise the safety property does not hold.

III. NILPOTENT

In this section, we give a decision procedure to the reachability problem (15) in section II-C when all A_i s in (4) are nilpotent.

A. Reformation of the Problem

Given an SS as (4), the initial set X and the unsafe set Y are two semi-algebraic sets, all the matrices A_1, \ldots, A_m are nilpotent and every component of u_1, u_2, \ldots, u_m is a polynomial vector w.r.t. t, determine whether the following formula holds or not,

$$\mathscr{F}(\mathbf{X},\mathbf{Y}) = \exists \mathbf{x} \exists \mathbf{y} \exists t : \mathbf{x} \in \mathbf{X} \land \mathbf{y} \in \mathbf{Y} \land t \ge 0 \land \Phi(\mathbf{x},t) = \mathbf{y}.$$
(16)

B. The Solution Form

In this section, we compute the solution form of an SS of (4) in which all the matrices A_1, \ldots, A_m are nilpotent and all u_1, u_2, \ldots, u_m are polynomials, and show the solution is a polynomial in $\mathbb{R}[\mathbf{x},t]$ by induction on the number of blocks of variables.

We first prove it when m = 1, *i.e.* the linear case.

Lemma 1. Given a linear system $\dot{\xi} = A\xi + \mathbf{u}(t)$ satisfying $A \in \mathbb{R}^{n_1 \times n_1}$ is a nilpotent matrix, and $\mathbf{u}(t) \in \mathbb{R}[t]^{n_1}$, a given initial point $\mathbf{x} \in \mathbb{R}^{n_1}$, the solution $\Phi(\mathbf{x},t)$ of the linear system is a polynomial in $\mathbb{R}[\mathbf{x},t]$.

Proof. Clearly, in this case,

$$\Phi(\mathbf{x},t) = e^{At}\mathbf{x} + \int_0^t e^{A(t-\tau)}\mathbf{u}(\tau)d\tau.$$

Since $A \in \mathbb{R}^{d \times d}$ is a nilpotent matrix, $A^k = 0$ for any $k \ge d$. Thus, $e^{At} = \sum_{k=0}^{d-1} \frac{t^k}{k!} A^k$. Moreover,

$$\Phi(\mathbf{x},t) = \sum_{k=0}^{d-1} \frac{t^k}{k!} A^k \mathbf{x} + \int_0^t \left(\sum_{k=0}^{d-1} \frac{(t-\tau)^k}{k!} A^k \mathbf{u}(\tau) \right) d\tau.$$

As A, \ldots, A^{d-1} are all real matrices in $\mathbb{R}^{n_1 \times n_1}$, it is easy to see that $\sum_{k=0}^{d-1} \frac{t^k}{k!} A^k \mathbf{x}$ is a polynomial vector in \mathbf{x} and t, and $\sum_{k=0}^{d-1} \frac{(t-\tau)^k}{k!} A^k u(\tau)$ is a polynomial vector in \mathbf{x} , t and τ . Hence,

$$\int_0^t \left(\sum_{k=0}^{d-1} \frac{(t-\tau)^k}{k!} A^k \mathbf{u}(\tau) \right) d\tau$$

is a polynomial in **x** and *t*. Thus,

$$\sum_{k=0}^{d-1} \frac{t^k}{k!} A^k \mathbf{x} + \int_0^t \left(\sum_{k=0}^{d-1} \frac{(t-\tau)^k}{k!} A^k \mathbf{u}(\tau) \right) d\tau$$

Theorem 3. Given an SS as (4) in which all the matrices A_1, \ldots, A_m are nilpotent and all $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m$ are polynomial vectors, then for a given initial point **x**, the solution $\Phi(\mathbf{x},t)$ is a polynomial vector in $\mathbb{R}[\mathbf{x},t]^n$, where $n = n_1 + \cdots + n_m$.

Proof. Let $\mathbf{x} = (\mathbf{z}_1, \dots, \mathbf{z}_m)$ correspond to $(\zeta_1, \dots, \zeta_m)$ in (4). For ζ_1 , by Lemma 1, we know $\zeta_1(t, \mathbf{x}) \in \mathbb{R}[\mathbf{x}, t]^{n_1}$.

Now, suppose that $\zeta_1(t, \mathbf{x}), \dots, \zeta_{k-1}(t, \mathbf{x})$ are all polynomial vectors in $\mathbb{R}[\mathbf{x},t]^{n_1},\ldots,\mathbb{R}[\mathbf{x},t]^{n_{k-1}}$, respectively, for $k \leq m$. We prove that $\zeta_k \in \mathbb{R}[\mathbf{x},t]^{n_k}$. Since $\mathbf{u}_k(t,\zeta_1,\ldots,\zeta_{k-1})$ is a polynomial vector, substituting $\zeta_1(t, \mathbf{x}), \ldots, \zeta_{k-1}(t, \mathbf{x})$ for $\zeta_1, \ldots, \zeta_{k-1}$ in $\mathbf{u}_k(t, \zeta_1, \ldots, \zeta_{k-1})$, it follows $\mathbf{u}_k(t, \zeta_1, \dots, \zeta_{k-1}) = \mathbf{u}_k(t, \mathbf{x}) \in \mathbb{R}[\mathbf{x}, t]^{n_k}$. Thus, the subdynamical system w.r.t. ζ_k is reduced to $\zeta_k = A_k \zeta_k + \mathbf{u}_k(t, \mathbf{x})$. By Lemma 1, this implies that $\zeta_k(t, \mathbf{x})$ is a polynomial vector. All in all, the solution $\Phi(\mathbf{x},t) \in \mathbb{R}[\mathbf{x},t]^n$.

Thus, (16) becomes decidable according to the decidability of Tarski algebra [39]. I.e.,

Theorem 4. The problem (16) is decidable.

IV. REAL EIGENVALUES

In this section, we give a decision procedure to the problem (15)when all A_i s are only with real eigenvalues and all \mathbf{u}_i s are PEF vectors in (4).

A. Reformulation of the Problem

Given an SS as (4), the initial set X and the unsafe set Y are two semi-algebraic sets, all the matrices A_1, \ldots, A_m have only real eigenvalues and $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m$ are PEF vectors, determine whether the following formula holds,

$$\mathscr{F}(\mathbf{X},\mathbf{Y}) = \exists \mathbf{x} \exists \mathbf{y} \exists t : \mathbf{x} \in \mathbf{X} \land \mathbf{y} \in \mathbf{Y} \land t \ge 0 \land \Phi(\mathbf{x},t) = \mathbf{y}.$$

B. Reduction to the Decision Problem of \mathcal{T}_e

In this part, we prove that the reachability problem above can be reduced to the decision problem of \mathcal{T}_e , therefore is decidable under the assumption of Schanuel's conjecture according to Strzeboński's result in [38].

Before proving the solution of (4) in this case can be represented as PEFs, we first show some properties on PEFs.

Lemma 2. The set of PEFs is closed under add, subtract, multiply and integral operations.

Proof. It is easy to see that the set of PEFs is closed under add, subtract and multiply operations. For the integral operation, since

$$\int e^{\lambda t} dt = \frac{1}{\lambda} e^{\lambda t}, \quad \text{and}$$

$$\int t^n e^{\lambda t} dt = \frac{t^n}{\lambda} e^{\lambda t} - \frac{nt^{n-1}}{\lambda^2} e^{\lambda t} + \dots + (-1)^n \frac{n(n-1)\cdots 1t}{\lambda^{n+1}} e^{\lambda t},$$
e integral of a PEF is still PEF.

the integral of a PEF is still PEF.

Lemma 3. Let $A \in \mathbb{R}^{n \times n}$ has real eigenvalues only, then e^{At} is a matrix with dimension $n \times n$, and all entries of e^{At} are PEFs.

Proof. Let J be the Jordan normal form of A, so there exist an invertible matrix Q such that $A = QJQ^{-1}$. Then, it follows

$$e^{At} = Qe^{Jt}Q^{-1}$$

Let

where J_1, J_2, \ldots, J_m be the corresponding Jordan blocks. Then

$$e^{At} = Qe^{Jt}Q^{-1} = Q\begin{bmatrix} e^{J_1t} & & \\ & e^{J_2t} & \\ & & \ddots & \\ & & & e^{J_mt} \end{bmatrix} Q^{-1}.$$

Without loss of generality, we just need to prove that all the elements of $e^{J_1 t}$ are PEFs. Suppose that the dimension of J_1 is $d \times d$ and the diagonal entry is λ , *i.e.*

$$J_1 = \lambda I + \begin{bmatrix} 0 & 1 & & \\ & 0 & \ddots & \\ & & \ddots & 1 \\ & & & & 0 \end{bmatrix}.$$

Denote the second summand of J_1 by M, obviously $M^d = 0$. So, we have

$$e^{J_{1}t} = e^{\lambda tI} \cdot e^{Mt} \\ = e^{\lambda tI} \cdot (I + tM + \frac{t^{2}}{2}M^{2} + \dots + \frac{t^{d-1}}{(d-1)!}M^{d-1}) \\ = \begin{bmatrix} e^{\lambda t} & & \\ & e^{\lambda t} & \\ & & \ddots & \\ & & & e^{\lambda t} \end{bmatrix} \cdot \begin{bmatrix} 1 & t & \cdots & \frac{t^{d-1}}{(d-1)!} \\ 1 & t & \cdots & \\ & & \ddots & \\ & & & 1 \end{bmatrix}.$$

Hence all the entries of $e^{J_1 t}$ are PEFs, and so are all entries of e^{At} .

Theorem 5. Given an SS of (4) in which all A_i s are only with real eigenvalues and all \mathbf{u}_i s are PEF vectors, and an initial $\mathbf{x} \in \mathbb{R}^n$, then its solution $\Phi(\mathbf{x},t)$ can be represented as of the following form

$$(\Phi(\mathbf{x},t))_i = \sum_{j=1}^{s_i} \phi_{ij}(\mathbf{x},t) e^{\mathbf{v}_{ij}t},$$
(17)

for i = 1, ..., n, where $\phi_{ij}(\mathbf{x}, t) \in \mathbb{R}[\mathbf{x}, t]$, $J_i \in \mathbb{N}$ and $v_{ij} \in \mathbb{R}$ for i = 1, ..., n, $j = 1, ..., s_i$.

Proof. Let $\mathbf{x} = (\mathbf{z}_1, \dots, \mathbf{z}_m)$ corresponding to $(\zeta_1, \dots, \zeta_m)$ in (4).

We proceed the proof by induction on m.

When m = 1, thus the solvable system (4) becomes a linear system. Whence the solution is

$$\zeta_1(t,\mathbf{x}) = e^{A_1 t} z_1 + \int_0^t e^{A_1(t-\tau)} u_1(\tau) d\tau$$

By Lemma 3, it follows that all entries in e^{A_1t} and $e^{A_1(t-\tau)}$ are PEFs. Moreover, using Lemma 2, we have $e^{A_1t}z_1 + \int_0^t e^{A_1(t-\tau)}u_1(\tau)d\tau$ is a PEF. Hence, $\zeta_1(t,\mathbf{x})$ is a PEF vector.

Now, suppose that $\zeta_1(t, \mathbf{x}), \ldots, \zeta_{k-1}(t, \mathbf{x}), k < m$, are all PEF vectors, we prove that $\zeta_k(t, \mathbf{x})$ is also a PEF vector. Since $\mathbf{u}_k(t, \zeta_1, \ldots, \zeta_{k-1})$ and $\zeta_1, \ldots, \zeta_{k-1}$ are all PEF vectors, substituting $\zeta_1(t, \mathbf{x}), \ldots, \zeta_{k-1}(t, \mathbf{x})$ for $\zeta_1, \ldots, \zeta_{k-1}$ in $\mathbf{u}_k(t, \zeta_1, \ldots, \zeta_{k-1})$, it follows $\mathbf{u}_k(t, \zeta_1, \ldots, \zeta_{k-1}) = \mathbf{u}_k(t, \mathbf{x})$ is a PEF vector. Thus, the sub-dynamical system w.r.t. ζ_k is reduced to

$$\zeta_k = A_k \zeta_k + \mathbf{u}_k(t, \mathbf{x}).$$

From the basis case, this implies that $\zeta_k(t, \mathbf{x})$ is a PEF vector.

In a word, the solution $\Phi(\mathbf{x},t)$ is a PEF vector, *i.e.* each of its component is of the following form

$$(\Phi(\mathbf{x},t))_i = \sum_{j=1}^{s_i} \phi_{ij}(\mathbf{x},t) e^{\mathbf{v}_{ij}t},$$
(18)

for i = 1, ..., n, where $\phi_{ij}(\mathbf{x}, t) \in \mathbb{R}[\mathbf{x}, t]$, $J_i \in \mathbb{N}$ and $v_{ij} \in \mathbb{R}$ for i = 1, ..., n, $j = 1, ..., s_i$.

Example 1.

$$\begin{bmatrix} \dot{\xi}_1 \\ \dot{\xi}_2 \\ \dot{\xi}_3 \end{bmatrix} = \begin{bmatrix} \xi_1 \\ \xi_1 - \xi_2 + e^t \\ -\xi_3 + \xi_1^2 \end{bmatrix},$$

with an initial state $\mathbf{x} = (x_1, x_2, x_3) \in \mathbb{R}^3$, the corresponding solution is

$$\begin{aligned} \xi_1(t, \mathbf{x}) &= e^t x_1, \\ \xi_2(t, \mathbf{x}) &= \left(\frac{x_1}{2} + \frac{1}{2}\right) e^t - \left(\frac{x_1}{2} + \frac{1}{2} - x_2\right) e^{-t}, \\ \xi_3(t, \mathbf{x}) &= \frac{x_1^2}{3} e^{2t} - \left(\frac{x_1^2}{3} - x_3\right) e^{-t}, \end{aligned}$$

which are PEFs.

Since X and Y are two semi-algebraic sets, there exist polynomials $p_1(\mathbf{x}), \ldots, p_J(\mathbf{x})$ such that

$$\mathbf{X} = \{ \mathbf{x} \in \mathbb{R}^n \mid p_1(\mathbf{x}) \rhd 0, \dots, p_{J_1}(\mathbf{x}) \rhd 0 \},\$$

$$\mathbf{Y} = \{ \mathbf{x} \in \mathbb{R}^n \mid p_{J_1+1}(\mathbf{x}) \rhd 0, \dots, p_J(\mathbf{x}) \rhd 0 \},\$$

where $\triangleright \in \{\geq, >\}$. Then (15) can be reduced to verify whether

$$\mathscr{F}(\mathbf{X},\mathbf{Y}) = \exists \mathbf{x} \exists \mathbf{y} \exists t : \Omega \tag{19}$$

holds, where,

$$\Omega = p_1(\mathbf{x}) \rhd 0 \land \dots \land p_{J_1}(\mathbf{x}) \rhd 0 \land p_{J_1+1}(\mathbf{y}) \rhd 0 \land \dots \land p_J(\mathbf{y}) \rhd 0$$
$$\land t \ge 0 \land \bigwedge_{i=1}^n \mathbf{y}_i = \sum_{j=1}^{s_i} \phi_{ij}(\mathbf{x}, t) e^{v_{ij}t}.$$

C. Decision Procedure for \mathcal{T}_e

In this part, we give a decision procedure for \mathcal{T}_e based on cylindrical algebraic decomposition (CAD), due to Collins [10].

The basic idea of CAD is: given a set S of polynomials in $\mathbb{R}[\mathbf{x}]$, CAD is used to partition \mathbb{R}^n into connected semi-algebraic sets, called cells, such that each polynomial in S keeps constant sign (either +, - or 0) on each cell. As CAD plays a fundamental role in computer algebra and real algebraic geometry, in the literature, a numerous works are done on improvement of CAD, e.g., [29], [22], [11], [13], [6], [19]. When constraints are open sets, GCAD [36] or OpenCAD [19] is enough, which partitions the space \mathbb{R}^n into a set of *open cells* instead of cells (i.e., takes sample points from open cells only), such that on each of which every polynomial in S keeps constant nonzero sign (either + or –). For example, suppose $f_1 = y - x$, $f_2 = y + x$. The graphs of $f_1 = 0$ and $f_2 = 0$ decompose \mathbb{R}^2 into 9 cells with different dimensions: four of which are 2-dimensional (open) cells (i.e., $f_1 \sim$ $0 \wedge f_2 \sim 0$, where $\sim \in \{>, <\}$); four of which are 1-dimensional cells (i.e., $f_1 \sim 0 \land f_2 = 0$, $f_1 = 0 \land f_2 \sim 0$, where $\sim \in \{>, <\}$); and one of which is 0-dimensional cell (i.e., $f_1 = 0 \land f_2 = 0$). Complete CAD takes at least one sample point from each of the 9 cells, while GCAD or openCAD takes at least one sample point only from each of the four 2-dimensional (open) cells. Formally,

Definition 3. For a polynomial $f(x_1,...,x_n) \in \mathbb{R}[x_1,...,x_n]$, a CAD (openCAD) defined by f under the order $x_1 \prec x_2 \prec \cdots \prec x_n$ is a set of sample points in \mathbb{R}^n obtained through the following three phases:

- Projection: Apply CAD (openCAD) projection operator on f to get a set of projection polynomials $\{f_n = f(x_1,...,x_n), f_{n-1}(x_1,...,x_{n-1}), ..., f_1(x_1)\};$
- Base: Choose a rational point in each of the (open) intervals defined by the real roots of f_1 ;
- Lifting: Substitute each sample point in \mathbb{R}^{i-1} for $(x_1, ..., x_{i-1})$ in f_i to get a univariate polynomial $f'_i(x_i)$, and then, as in Base phase, choose sample points for $f'_i(x_i)$. Repeat this process for i from 2 to n.

Using CAD (openCAD), we develop a decision procedure for \mathcal{T}_e as follows:

- Step 1 Check whether $X \cap Y = \emptyset$, if not, it's easy to see that (19) holds.
- Step 2 Translate the problem to an openCAD solvable problem if X and Y are open sets, otherwise a CAD solvable problem. By (18), $y_i(\mathbf{x},t) = \sum_{j=1}^{s_i} \phi_{ij}(\mathbf{x},t)e^{\mathbf{v}_{ij}t}$. So, we can replace $p_j(\mathbf{y})$ with $p_j(\mathbf{y}(\mathbf{x},t))$, which is polynomial in \mathbf{x} and polynomial-exponential in t, abbreviated as $p_j(\mathbf{x},t)$, for $j = J_1 + 1, \dots, J$. Simply, we define $p_j(\mathbf{x},t)$ as $p_j(\mathbf{x})$, for $j = 1, \dots, J_1$. Thus, $\mathscr{F}(\mathbf{X}, \mathbf{Y})$ in (19) can be reformulated as $\mathscr{F} = \exists \mathbf{x} \exists t \bigwedge_{j=1}^J p_j(\mathbf{x},t) \triangleright 0 \land t \ge 0$.
- Step 3 Eliminate x_1, \dots, x_n one by one using CAD (open-CAD) projection operator on $\prod_{j=1}^{J} p_j$ and obtain a set of projection polynomials $\{q_n(x_1, \dots, x_n, t) = \prod_{j=1}^{J} p_j, q_{n-1}(x_2, \dots, x_n, t)\}, \dots, q_0(t)\}$.
- Step 4 Isolate the real roots of the resulted PEF q_0 based on *Rolle's theorem*, which will be elaborated in the next subsection.
- Step 5 Lift the solution using openCAD or CAD lifting procedure corresponding to Step 2 according to the order t, x_n, \dots, x_1 based on $\{q_0, \dots, q_n\}$, and obtain a set S of sample points. Step 6 Check if \mathscr{F} holds by testing if there exists α in S such that
- $\wedge_{j=1}^{J} p_j(\alpha) \triangleright 0.$

In [38], Strzeboński presented another decision procedure for \mathcal{T}_e completely based on CAD. Our decision procedure differentiates from Strzeboński's in the following points:

- When all constraints are open sets, our method is based on openCAD, which requires less computation compared to the corresponding complete CAD, as we do not need to consider the cells that are represented as roots of equations involving polynomial-exponential functions, which are extremely difficult, during the base and lifting phases in openCAD. Therefore, as indicated later in the experiments, our decision procedure is more efficient in this case. But the two decision procedures share the same complexity in general case.
- In [38], an algorithm for isolating real roots of a given PEF based on *weak Fourier sequence* [37] is given. It is claimed that the algorithm is *complete* under the assumption of *Schanuel's conjecture* [34]. While, in this paper, we give another algorithm to isolate real roots of the resulted PEF $q_0(t)$ based on *Rolle's theorem*. We prove that our approach is also *complete* under the assumption that $q_0(t)$ does not have any multiple real roots, which can be implied by *Schanuel's conjecture*.

D. Isolating Real Roots of PEFs

In this part we give an algorithm PEFIsolation to isolate the finitely many real roots of a PEF.

Definition 4. Consider a PEF in t as

$$f(t) = \sum_{i=0}^{s} f_i(t) e^{\mathbf{v}_i t},$$
(20)

where $s \in \mathbb{N}$, $0 \not\equiv f_i \in \mathbb{R}[t]$ and $v_i \in \mathbb{R}$ are pairwise different. Real root isolation of the equation f(t) = 0 is to obtain a set of intervals

 $\{I_j = (a_j, b_j) \mid a_j, b_j \in \mathbb{R} \land a_j < b_j, j = 1, ..., J\}$ such that $I_i \cap I_j = \emptyset$ if $i \neq j$, in each I_j there exists only one real root of f(t), and all real roots of f(t) are contained in $\bigcup_{i=1}^J I_j$.

Given an open interval *I*, real root isolation of f(t) over *I* can be defined similarly.

Without loss of generality, in (20), we can assume

$$0 = v_0 < v_1 < v_2 < \dots < v_s, f_i(t) \neq 0, \text{ for } i = 0, 1, \dots, s, \quad (21)$$

since we can always multiply out by $e^{v_0 t}$ for the smallest v_0 to ensure this happens. When s = 1 or every v_i ($0 \le i \le s$) is a positive integer, in [2] an algorithm named ISOL was proposed to isolate all real roots of f(t). This algorithm can be easily extended to the case when all v_i ($i = 0, \dots, s$) are rationals or there exists a nonzero real number κ such that for every $0 \le i \le s$, $v_i \kappa$ is a rational.

1) Lower and Upper Bounds on Real Roots

Similar to [2], we can prove the following theorem, which indicates that there is a lower and upper bound on real roots for any given PEF.

Theorem 6 (upper bound). Let f(t) be a PEF of the form (20). Then we can obtain an upper bound C on its real roots through the following procedure:

- 1) Find $C_1 \ge 0$, M > 0 such that for all $t > C_1$, $|f_s(t)| > \frac{1}{M}$;
- 2) Find $C_2 \ge 0$ and $k \in \mathbb{N}$ such that for all $t > C_2$ and for all $0 \le i < s$, $|f_i(t)| < \frac{t^k}{sM}$;
- 3) Find $C_3 \ge 0$ such that for all $t > C_3$, $t^k < e^{(v_s v_{s-1})t}$; 4) Set $C = \max\{C_1, C_2, C_3\}$.

Proof. Let t > C, then we have $|f_s(t)| > \frac{1}{M}$, $t^k < e^{(v_s - v_{s-1})t}$, $|f_i(t)| < \frac{t^k}{M}$, for $i = 0, \dots, s-1$. Whence

$$\begin{aligned} |f_0(t) + \sum_{i=1}^{s-1} f_i(t) e^{v_i t}| &\leq |f_0(t)| + \sum_{i=1}^{s-1} |f_i(t) e^{v_i t}| < \frac{t^k}{sM} + \sum_{i=1}^{s-1} \frac{t^k}{sM} e^{v_i t} \\ &< \frac{t^k}{sM} e^{v_{s-1} t} + \sum_{i=1}^{s-1} \frac{t^k}{sM} e^{v_{s-1} t} = \frac{1}{M} t^k e^{v_{s-1} t} < \frac{1}{M} e^{v_s t} < |f_s(t) e^{v_s t}|. \end{aligned}$$

Thus, $|f_0(t) + \sum_{i=1}^{s-1} f_i(t)e^{v_i t}| < |f_s(t)e^{v_s t}|$, and we have $f_0(t) + \sum_{i=1}^{s-1} f_i(t)e^{v_i t} + f_s(t)e^{v_s t} \neq 0$. This implies $f(t) \neq 0$ for any $t \ge C$. So C is an upper bound on the real roots of f(t).

In order to get a lower bound, a commonly used method is to replace f(t) with $g(t) = f(-t)e^{v_s t}$. Then, by Theorem 6, there is an upper bound *B* on the real roots of g(t) = 0. It's easy to see that -B is a lower bound on the real roots of f(t) = 0. Thus, we see that all roots of f(t) = 0 are in the interval (-B, C). In what follows, we denote by L(f) = -B, U(f) = C, the lower and upper bounds on the real roots of f(t), respectively.

2) Algorithm

In this part, we present our algorithm PEFIsolation for isolating all real roots of a given nonzero PEF f(t) of the form (20).

Definition 5. Let f(t) be a nonzero PEF of the form (20), then we define

$$\operatorname{coff}(f) = (f_0, f_1, \dots, f_s)^T, \operatorname{nu}(f) = (0, v_1, \dots, v_s)^T,$$

 $\operatorname{deg}(f) = (\operatorname{deg}(f_0), \operatorname{deg}(f_1), \dots, \operatorname{deg}(f_s))^T,$

where deg(g) means the degree of g, and as a convention, deg(0) = -1. So, (20) can be shorten as

$$f(t) = \operatorname{coff}(f)^T \cdot e^{\operatorname{nu}(f)t},$$

where $e^{\operatorname{nu}(f)t} = (1, e^{v_1 t}, \dots, e^{v_s t})^T$, $\mathbf{a} \cdot \mathbf{b}$ stands for the inner product of the two vectors, i.e., $\sum_{i=1}^n a_i b_i$.

 $\operatorname{coff}(f') = (f'_0, f'_1 + v_1 f_1(t), \dots, f'_s + v_s f_s(t))^T, \ \operatorname{nu}(f') = (0, v_1, \dots, v_s)^T, \\ \operatorname{deg}(f') = (\max\{\operatorname{deg}(f_0) - 1, -1\}, \ \operatorname{deg}(f_1), \dots, \operatorname{deg}(f_s))^T,$

where f' denotes the derivative of f w.r.t. t.

In the following, we will explain the basic idea behind PEFIsolation through the following simple example.

Example 2. Consider
$$\hat{f}(t) = t + 1 + e^{\sqrt{2t}} - (t+2)e^{\sqrt{5t}}$$

Firstly, in order to isolate the real roots of $\hat{f}(t) = 0$, we need to calculate the upper and lower bounds on all its real roots according to Theorem 6.

Regarding the upper bound of $\hat{f}(t) = 0$, we have: (i) $C_1 = 0, M = 1$, $\forall t \ge 0, |t+2| > 1$; (ii) $C_2 = 4, k = 2, \forall t \ge 4, |t+1| < \frac{t^2}{2}, 1 < \frac{t^2}{2}$; (iii) $C_3 = 12, \forall t \ge 12, t^2 < e^{(\sqrt{5} - \sqrt{2})t}$. Thus, we obtain $U(\hat{f}) = 12$. In order to obtain the lower bound, we have to calculate the upper

In order to obtain the lower bound, we have to calculate the upper bound U(g) of $g(t) = \hat{f}(-t)e^{\sqrt{5}t}$, *i.e.*, $g(t) = t - 2 + e^{(\sqrt{5} - \sqrt{2})t} - (t - 1)e^{\sqrt{5}t}$. Because (i) $\overline{C}_1 = 3$, M = 1, $\forall t \ge 3$, |t - 1| > 1; (ii) $\overline{C}_2 = 4$, k = 2, $\forall t \ge 4$, $|t - 2| < \frac{t^2}{2}$ and $1 < \frac{t^2}{2}$; (iii) $\overline{C}_3 = 1$, $\forall t \ge 1$ and $t^2 < e^{\sqrt{2}t}$, we obtain the upper bound U(g) = 4.

Therefore, the lower bound $L(\hat{f}) = -U(g) = -4$ is obtained. Obviously, all real roots of $\hat{f}(t) = 0$ should be in the interval (-4, 12), which implies that we just need to isolate all real roots in (-4, 12).

From differential mean value theorem (i.e., Rolle's theorem), we know there must exist at last one real root of f'(t) = 0 between every two real roots of f(t) = 0, if f(t) is continuous differentiable. In order to obtain the real roots of f(t) = 0, we can try to get the real roots of f'(t) = 0 first. Likewise, in order to obtain the real roots of f'(t) = 0, we can try to get the real roots of f''(t) = 0 first. We can repeat the above procedure until the real solutions of the *i*th derivative of f(t) for some *i* can be achieved. Then, we lift the real solutions of the respective derivative in the inverse order until f(t) itself. We illuminate the procedure by continuing the running example.

At the beginning,

$$S_0 = \hat{f}(t) = t + 1 + e^{\sqrt{2}t} - (t+2)e^{\sqrt{5}t},$$

$$coff(S_0) = (t+1, 1, -t-2)^T,$$

$$nu(S_0) = (0, \sqrt{2}, \sqrt{5})^T, deg(S_0) = (1, 0, 1)^T.$$

Then, we obtain the derivative of \hat{f} is

$$\begin{split} S_1 &= \hat{f}'(t) = 1 + \sqrt{2}e^{\sqrt{2}t} - (\sqrt{5}t + 2\sqrt{5} + 1)e^{\sqrt{5}t}, \\ \mathrm{coff}(S_1) &= (1, \sqrt{2}, -\sqrt{5}t - 2\sqrt{5} - 1)^T, \\ \mathrm{nu}(S_1) &= (0, \sqrt{2}, \sqrt{5})^T, \mathrm{deg}(S_1) = (0, 0, 1)^T. \end{split}$$

Furthermore, the derivative of \hat{f}' is

$$\hat{f}''(t) = 0 + 2e^{\sqrt{2}t} - (5t + 2\sqrt{5} + 10)e^{\sqrt{5}t}, \\ \operatorname{coff}(\hat{f}'') = (0, 2, -5t - 2\sqrt{5} - 10)^T, \\ \operatorname{nu}(\hat{f}'') = (0, \sqrt{2}, \sqrt{5})^T, \operatorname{deg}(\hat{f}'') = (-1, 0, 1)^T.$$

Clearly, \hat{f}'' and the following S_2 share the same real roots:

$$S_{2} = \hat{f}''(t)e^{-\sqrt{2}t} = 2 - (5t + 2\sqrt{5} + 10)e^{(\sqrt{5} - \sqrt{2})t}, \qquad (22)$$

$$\operatorname{coff}(S_{2}) = (0, 2, -5t - 2\sqrt{5} - 10)^{T}, \qquad (23)$$

$$\operatorname{nu}(S_{2}) = (0, 0, \sqrt{5} - \sqrt{2})^{T}, \operatorname{deg}(S_{2}) = (-1, 0, 1)^{T}.$$

Now, the derivative of S_2 is

$$S_3 = S'_2 = 0 + 0 + he^{(\sqrt{5} - \sqrt{2})t}, \text{coff}(S_3) = (0, 0, h)^T, \text{nu}(S_3) = (0, 0, \sqrt{5} - \sqrt{2})^T, \text{deg}(S_3) = (-1, -1, 1)^T.$$

where $h = -(5(\sqrt{5} - \sqrt{2})t + 15 + 10\sqrt{5} - 2\sqrt{10} - 10\sqrt{2})$. Obviously, $S_3 = 0$ if and only if h = 0, while the real zeros of h can be easily achieved by any real root isolation procedure for polynomials [12].

Remark 1. In general, suppose $S_i(t) = f_0(t) + \sum_{j=1}^J f_j(t)e^{v_j t}$ with $0 \neq f_j(t) \in \mathbb{R}[t], 0 < v_1 < \cdots < v_J$, and $0 < J \in \mathbb{N}$, then we define $S_{i+1}(t) = S'_i(t)$ if $f'_0(t) \neq 0$; otherwise, $S_{i+1}(t) = S'_i(t)e^{-v_1 t} = (f'_1(t) + v_1f_1(t)) + \sum_{j=2}^J (f'_j(t) + v_jf_j(t))e^{(v_j - v_1)t}$. It's obvious that $S_{i+1} = 0$ shares the same real roots of $S'_i(t) = 0$. We construct S_{i+1} from S_i ,

for $i = 0, \dots$. This procedure terminates when S_k is a polynomial for some k.

Theorem 7. Let f(t) be a PEF, f'(t) the derivative of f(t) w.r.t. t, I = (a,b) a non-empty open interval, and $\mathcal{L}_I(f') = \{I_j | j = 1, ..., J\}$ a real root isolation of f' in I, in which $I_j = (a_j, b_j)$ with $a = b_0 < a_1 < b_1 < \cdots < a_J < b_J < a_{J+1} = b$. Furthermore, f(t) has no real roots in any closed interval $[a_j, b_j]$, $1 \le j \le J$. Then, $\{ (b_j, a_{j+1}) \mid f(b_j)f(a_{j+1}) < 0, \ 0 \le j \le J \}$ is a real root isolation of f(t) in I.

Proof. Since f(t) has no real roots in any closed interval $[a_j, b_j]$, $1 \le j \le J$, all real roots of f(t) are in $\bigcup_{j=0}^{J} (b_j, a_{j+1})$ and $f(b_j)f(a_{j+1}) \ne 0$. Moreover, f(t) has at most one real root in each (b_j, a_{j+1}) , otherwise, there must be at least one real root of f'(t) = 0 on it by *Rolle's theorem*, which is a contradiction with the definition of $\mathscr{L}_I(f')$. So, if $f(b_j)f(a_{j+1}) < 0$ then there exists only one real root of f(t) in (b_j, a_{j+1}) , otherwise no real root of f(t) in (b_j, a_{j+1}) . This completes the proof.

Now, let's continue the running example. As $e^{(\sqrt{5}-\sqrt{2})t} \neq 0$, by $S_3 = he^{(\sqrt{5}-\sqrt{2})t} = 0$, it follows h(t) = 0. Thus, $t = -\frac{15+10\sqrt{5}-2\sqrt{10}-10\sqrt{2}}{5(\sqrt{5}-\sqrt{2})} \in (-5,-4)$. As $(-5,-4) \cap (-4,12) = \emptyset$, there is no real root of $S_3 = 0$ in (-4,12). Hence, we have $\mathscr{L}(S_3) = \emptyset$. In addition, from (22), we have

$$S_2(-4) = 2 + (10 - 2\sqrt{5})e^{-4(\sqrt{5} - \sqrt{2})} > 0$$

$$S_2(12) = 2 - (70 + 2\sqrt{5})e^{12(\sqrt{5} - \sqrt{2})} < 0.$$

So, there exists only one real root of S_2 in (-4, 12) by Theorem 7. Clearly, the real root isolation of S_2 in (-4, 12) is same as that of \hat{f}'' .

In order to construct $\mathscr{L}_{(-4,12)}(S_1)$, a real root isolation of S_1 in (-4,12), from $\mathscr{L}_{(-4,12)}(S_2)$ by Theorem 7, the condition that there is no real root of S_1 in [a,b] for any (a,b) in $\mathscr{L}_{(-4,12)}(S_2)$ should be guaranteed. This means that we have to refine the intervals in $\mathscr{L}_{(-4,12)}(S_2)$ until the condition holds. This is achieved by Algorithm 2 below (see lines 2-13).

The following table is the bisection procedure (line 2-13) in Algorithm 2 to refine the interval (-4, 12), in which ' \exists ' (resp. ' \neg \exists ') means there exists (no) a real root in the observed interval.

	(-4,12)	(-4,4)	(-4,0)	(-2,0)	(-2,-1)
<i>S</i> ₂	Ш	Ξ	Ξ	Ξ	П
S_1	Ξ	Ξ	П	П	LL L

Finally, a refined interval (a,b) = (-2,-1) is obtained, which satisfies the condition of Theorem 7. Thus, (-4,-2) and (-1,12)are two intervals that may contain at most one real root of $S_1(t) = 0$. In addition, as $S_1(-4)S_1(-2) > 0$ and $S_1(-1)S_1(12) < 0$, (-1,12)contains a real root of $S_1(t) = 0$, but (-4,-2) does not by Theorem 7. Thus, we get a real root isolation for $S_1(t) = 0$ in (-4,12), i.e., $\mathscr{L}_{(-4,12)}(S_1) = \{(-1,12)\}.$

In order to compute $\mathscr{L}_{(-4,12)}(S_0)$, we repeat the above procedure, and obtain $\mathscr{L}_{(-4,12)}(\hat{f}) = \{(-4,-0.59375),(-0.390625,12)\}.$

Up to now, we have already explained the main idea of our approach how to isolate real roots of a PEF by the running example. This procedure is implemented in Algorithm 1, whose main steps can be understood as follows:

- Step 1: In line 1, compute upper and lower bounds of the real roots of f(t);
- Step 2: In line 3, construct a sequence $S_0(t) = f(t)$, $S_1(t)$, $S_2(t)$, ..., $S_r(t)$, where S_i is a PEF which has the same real roots as the derivative of S_{i-1} , $i = 1, 2, ..., r, r \in \mathbb{N}$, and $S_r(t)$ is a polynomial in t.

- Step 3: Isolate all real roots of $S_r(t)$ by calling UPIsolating $(S_r(t))$ in line 4. Note that the problem of isolating real roots of a univariate polynomial is well studied (e.g. in [12]).
- Step 4: In line 6, for i = r 1 down to 0, construct a real root isolation of S_i from that of S_{i+1} using Theorem 7 by calling PEFI. Note that during this procedure, we use \mathscr{I}_1 to record all subintervals in which f(t) has no real roots, while \mathscr{I}_2 to record all subintervals in which f'(t) has no real roots. So, we only need to construct a real root isolation of S_i from that of S_{i+1} on the remainder part of the considered interval by excluding all subintervals in \mathscr{I}_1 and \mathscr{I}_2 , and accordingly update \mathscr{I}_1 and \mathscr{I}_2 in each iteration, see the detail in Algorithm 2.

Algorithm 1: PEFIsolation

- **Input:** f(t), a PEF of the form (20) with the assumption (21), which has no multiple real roots **Output:** \mathscr{L} , a real root isolation of f(t)
- 1 Calculate a lower bound a and an upper bound b on real roots of f(t) = 0;
- 2 set $\mathscr{I}_1 \leftarrow \emptyset, \mathscr{I}_2 \leftarrow \emptyset;$
 - /* \mathscr{I}_1 records all closed subintervals of [a,b] in which f(t) has no real roots, while \mathscr{I}_2 records all closed subintervals of [a,b] in which f'(t) has no real roots. */
- 3 Construct a sequence, $S_0(t) = f(t)$, $S_1(t)$, $S_2(t)$, ..., $S_r(t)$, where S_i is a PEF, which shares the common real roots with the derivative of S_{i-1} , i = 1, 2, ..., r, $r \in \mathbb{N}$, and $S_r(t)$ is a polynomial;
- 4 $\mathscr{L}_{(a,b)}(S_r) := \text{UPIsolating}(S_r(t))$, a real root isolation of $S_r(t)$;
- **5** for $i = r 1; i \ge 0; i -$ **do**
- $\mathbf{6} \quad \left[\begin{array}{c} [\mathscr{I}_1, \mathscr{I}_2, \mathscr{L}] \leftarrow \texttt{PEFI}(S_0, S_1, S_i, S_{i+1}, (a, b), \mathscr{I}_1, \mathscr{I}_2, \mathscr{L}); \end{array} \right]$
- 7 for $[c,d] \in \mathscr{I}_2$ do
- 8 | if $S_0(c)S_0(d) < 0$ then

```
10 return \mathscr{L};
```

Theorem 8 (Correctness of PEFI). Algorithm PEFI always terminates correctly.

Proof. The termination of PEFI is obvious because $f_1(t) = 0$ and $f_2(t) = 0$ have no common real roots. Then we prove its correctness.

 \mathscr{I}'_1 and \mathscr{I}'_2 are updated in line 5 and line 7, respectively. Obviously, after every update, the properties of \mathscr{I}'_1 and \mathscr{I}'_2 still hold, i.e., $f_1(t)$ has no real roots in $\cup \mathscr{I}'_1$, $f_2(t)$ has no real roots in $\cup \mathscr{I}'_2$, and $\cup \mathscr{I}'_1 \cap \cup \mathscr{I}'_2 = \emptyset$. It is also easy to see that, after the *for loop* at lines 15-18, \mathscr{L}' is a real root isolation of $g_1(t)$ on $(a,b) \setminus \cup (\mathscr{I}'_1 \cup \mathscr{I}'_2)$.

Theorem 9 (Correctness of PEFIsolation). Algorithm PEFIsolation always terminates and returns a real root isolation for a given PEF f, if f does not have multiple real roots.

Proof. Termination is immediately obtained from Theorem 8. Then we prove its correctness. After the *for loop* in line 2, \mathscr{L} is a real root isolation of $S_0(t) = 0$ (i.e., f(t) = 0) on $(a,b) \setminus \bigcup (\mathscr{I}_1 \cup \mathscr{I}_2)$. Because f'(t) has a constant nonzero sign in each interval of \mathscr{I}_2 , f(t) has at most one real root in each interval of \mathscr{I}_2 and this can be decided by checking the signs of f(t) at two endpoints of the interval. Moreover, since there is no real root of f(t) = 0 in $\bigcup \mathscr{I}_1$, so \mathscr{L}_2 is a real root isolation of $S_0(t)$ in (a,b).

Algorithm 2: PEFI

- **Input:** (1) PEFs $f_1(t), f_2(t), g_1(t), g_2(t)$ s.t. $f_2(t)$ and $f'_1(t)$ share same real zeros, $g_2(t)$ and $g'_1(t)$ share same real zeros, and $f_1(t)$ and $f_2(t)$ have no common real zeros; (2) an open interval (a,b);
- (3) $\mathscr{I}_1, \mathscr{I}_2$, two sets of closed intervals contained in (a, b), s.t. $f_1(t)$ has no real zeros in $\cup \mathscr{I}_1, f_2(t)$ has no real zeros in $\cup \mathscr{I}_2, \cup \mathscr{I}_1 \cap \cup \mathscr{I}_2 = \emptyset$;
- (4) \mathscr{L} , a real root isolation of $g_2(t)$ on $(a,b) \setminus \cup (\mathscr{I}_1 \cup \mathscr{I}_2)$.
- **Output:** (1) \mathscr{I}'_1 and \mathscr{I}'_2 with the same properties as \mathscr{I}_1 and \mathscr{I}_2 , respectively;

(2) \mathscr{L}_2 , a real root isolation of $g_1(t)$ on $(a,b) \setminus \cup (\mathscr{I}'_1 \cup \mathscr{I}'_2)$.

- $\mathbf{1} \ \mathscr{I}_1' \leftarrow \mathscr{I}_1, \ \mathscr{I}_2' \leftarrow \mathscr{I}_2;$
- 2 for (l,u) in \mathscr{L} do

3

4

- while $0 \in g_1([l,u])$ do if $0 \notin f_1([l,u])$ then
- $| \mathscr{I}'_1 \leftarrow \mathscr{I}'_1 \cup \{[l,u]\}; \text{ break};$ 5 if $0 \notin f_2([l,u])$ and $f_1(l)f_1(u) \neq 0$ then 6 $\mathscr{I}_{2}' \leftarrow \mathscr{I}_{2}' \cup \{[l, u]\}; \text{ break};$ 7 if $g_2(l)g_2(\frac{l+u}{2}) < 0$ then 8 $u \leftarrow \frac{l+u}{2};$ 9 else if $g_2(\frac{l+u}{2}) = 0$ then 10 $l \leftarrow \frac{3l+u}{4}; \quad u \leftarrow \frac{l+3u}{4};$ 11 else 12 $l \leftarrow \frac{l+u}{2};$ 13 14 $\mathscr{L}_1 \leftarrow \mathscr{L}; \ \mathscr{L}_2 \leftarrow \emptyset; \ \mathscr{L}_3 \leftarrow \{(a_1, b_1), \dots, (a_m, b_m)\};$ /* where $a_1, b_1, \cdots, a_m, b_m$ are the endpoints of the intervals in \mathscr{I}_1 and \mathscr{I}_1 s.t. $a \leq a_1 < b_1 < \cdots < a_m < b_m \leq b$, $(a_i, b_i) \subseteq (a, b) \setminus$ $\cup (\mathscr{I}_1 \cup \mathscr{I}_2)$ for $i=1,\ldots,$ m, and $\cup_{i=1}^{m}(a_i,b_i)=(a,b)\setminus \cup(\mathscr{I}_1\cup\mathscr{I}_2).$ 15 for (c,d) in \mathcal{L}_3 do
- 16 $\mathscr{L}_{1(c,d)} \leftarrow \{I \mid I \in \mathscr{L}_1 \text{ and } I \subset (c,d)\};$
- 17 Obtain a real root isolation $\mathscr{L}_{(c,d)}$ for $g_1(t)$ on (c,d) from $\mathscr{L}_{1(c,d)}$ by Theorem 7;

*/

- 18 $\[\mathcal{L}_2 \leftarrow \mathscr{L}_2 \cup \mathscr{L}_{(c,d)};\]$
- 19 return $\mathscr{I}'_1, \mathscr{I}'_2, \mathscr{L}_2;$

3) Multiple Real Roots of PEFs

Termination of the algorithm PEFIsolation with an input PEF f rely on that f does not have multiple real roots, which is however not obvious to check. In this section, we deal with multiple real roots of PEF based on Schanuel's conjecture.

Definition 6 (Algebraic independence). A set of complex numbers $S = \{a_1, ..., a_n\}$ is algebraically independent over \mathbb{Q} if the elements of *S* do not satisfy any non-trivial polynomial equation with coefficients in \mathbb{Q} .

Definition 7 (Transcendence degree). Let L be a field extension of \mathbb{Q} , the transcendence degree of L over \mathbb{Q} is defined as the largest cardinality of an algebraically independent subset of L over \mathbb{Q} .

Conjecture 1 (Schanuel's conjecture). *Given any complex numbers* z_1, \ldots, z_n that are linearly independent over \mathbb{Q} , the extension field $\mathbb{Q}(z_1, \ldots, z_n, e^{z_1}, \ldots, e^{z_n})$ has transcendence degree of at least n over \mathbb{Q} .

In what follows, we handle the multiple real roots of PEF. Let

$$f(t) = f_0(t) + f_1(t)e^{\lambda_1 t} + \dots + f_r(t)e^{\lambda_r t},$$

where $f_0, \ldots, f_r \in \mathbb{Q}[t]$, and $\lambda_1, \ldots, \lambda_r$ are different algebraic numbers. Let a_1, \ldots, a_n be an integer-basis of $\lambda_1, \ldots, \lambda_r$. Then a_1, \ldots, a_n are linearly independent over \mathbb{Q} , and f(t) is a polynomial w.r.t. $t, e^{a_1 t}, \ldots, e^{a_n t}$, denoted by $f(t, e^{a_1 t}, \ldots, e^{a_n t})$.

Since $f(t, y_1, \dots, y_n)$ is a polynomial, by factorization we have

$$f(t, y_1, \dots, y_n) = f_1^{m_1}(t, y_1, \dots, y_n) \cdots f_s^{m_s}(t, y_1, \dots, y_n),$$

whose square free part is denoted by

$$f(t,y_1,\ldots,y_n)=f_1(t,y_1,\ldots,y_n)\cdots f_s(t,y_1,\ldots,y_n).$$

This yields a PEF $\hat{f}(t, e^{a_1 t}, \dots, e^{a_n t})$, denoted as the square free part of $f(t, e^{a_1 t}, \dots, e^{a_n t})$.

The following corollaries can be derived based on Schanuel's conjecture:

Corollary 3. Let a_1, \ldots, a_n be algebraic numbers that are linearly independent over \mathbb{Q} . The transcendence degree of the field extension $\mathbb{Q}(t, e^{a_1 t}, \ldots, e^{a_n t})$ is at least n, if $t \neq 0$.

Proof. Since a_1, \ldots, a_n are linearly independent over \mathbb{Q} and $t \neq 0$, a_1t, \ldots, a_nt are linearly independent over \mathbb{Q} . By Schanuel's conjecture, the transcendence degree of the field extension

$$\mathbb{Q}(a_1t,\ldots,a_nt,e^{a_1t},\ldots,e^{a_nt})$$

is at least *n*. Besides, a_1, \ldots, a_n are algebraic numbers, thus $\mathbb{Q}(t) = \mathbb{Q}(a_1t, \ldots, a_nt)$, i.e., $\mathbb{Q}(a_1t, \ldots, a_nt, e^{a_1t}, \ldots, e^{a_nt}) = \mathbb{Q}(t, e^{a_1t}, \ldots, e^{a_nt})$. Therefore, The transcendence degree of the field extension $\mathbb{Q}(t, e^{a_1t}, \ldots, e^{a_nt})$ is at least *n*.

Corollary 4. Let $f(t, e^{a_1t}, \ldots, e^{a_nt})$ be a PEF w.r.t. t, and thus a polynomial w.r.t. $t, e^{a_1t}, \ldots, e^{a_nt}$, where a_1, \ldots, a_n are linearly independent. Suppose $f(t, y_1, \ldots, y_n)$ is square free, then $f(t, e^{a_1t}, \ldots, e^{a_nt})$ has no multiple real root except 0.

Proof. Since $f(t, y_1, \ldots, y_n)$ is square free, we may write

$$f(t, y_1, \dots, y_n) = f_1(t, y_1, \dots, y_n) \cdots f_m(t, y_1, \dots, y_n)$$

where, for any $1 \le i, j \le m, i \ne j, f_i(t, y_1, \dots, y_n)$ is irreducible and $f_i(t, y_1, \dots, y_n)$ and $f_j(t, y_1, \dots, y_n)$ are co-prime.

We first prove, by contradiction, that $f_i(t, e^{a_1t}, \dots, e^{a_nt})$ and $f_j(t, e^{a_1t}, \dots, e^{a_nt})$ have no nonzero common real root. Suppose $t_0 \neq 0$ is a common real root of $f_i(t, e^{a_1t}, \dots, e^{a_nt})$ and $f_j(t, e^{a_1t}, \dots, e^{a_nt})$. By Corollary 3, we have that the transcendence degree of $\mathbb{Q}(t_0, e^{a_1t_0}, \dots, e^{a_nt_0})$ is at least *n*. Then there must exist *n* elements in $\{t_0, e^{a_1t_0}, \dots, e^{a_nt_0}\}$ that are algebraically independent. Without loss of generality, let $\{t_0, e^{a_1t_0}, \dots, e^{a_{n-1}t_0}\}$ be the *n* elements that are algebraically independent. Besides, let $g(t, y_1, \dots, y_{n-1})$ be the resultant of $f_i(t, y_1, \dots, y_n)$ and $f_j(t, y_1, \dots, y_n)$. Further since $f_i(t, y_1, \dots, y_n)$ and $f_j(t, y_1, \dots, y_{n-1})$. Further since $f_i(t, y_1, \dots, y_n)$ and $f_j(t, y_1, \dots, y_{n-1})$ is non-trivial polynomial, indicating that $(t_0, e^{a_1t_0}, \dots, e^{a_{n-1}t_0})$ is a real root of some non-trivial polynomial. This contradicts that $\{t_0, e^{a_1t_0}, \dots, e^{a_{n-1}t_0}\}$ are algebraically independent. Consequently, $f_i(t, e^{a_1t}, \dots, e^{a_n})$ and $f_j(t, e^{a_1t_0}, \dots, e^{a_n})$ have no nonzero common real root.

Next, we prove that $f_i(t, e^{a_1 t}, \dots, e^{a_n t})$ has no multiple real root. Suppose

$$f_i(t, e^{a_1 t}, \dots, e^{a_n t}) = h_0(t) + \sum_{j=1}^s h_j(t) (e^{a_1 t})^{b_{j_1}} \cdots (e^{a_n t})^{b_{j_n}},$$

where, $h_0(t), \ldots, h_n(t)$ are non-trivial polynomials, $b_{jk} \in \mathbb{N}$, $1 \le j \le s$, and $1 \le k \le n$. Then we have

$$f'_{i}(t, e^{a_{1}t}, \dots, e^{a_{n}t}) = h'_{0}(t) + \sum_{j=1}^{s} (h'_{j}(t) + (a_{1}b_{j1} + \dots + a_{n}b_{jn})h_{j}(t))(e^{a_{1}t})^{b_{j1}} \cdots (e^{a_{n}t})^{b_{jn}}.$$

Moreover,

$$f_i(t, y_1, \dots, y_n) = h_0(t) + \sum_{j=1}^s h_j(t) y_1^{b_{j1}} \cdots y_n^{b_{jn}},$$

$$f'_i(t, e^{a_1 t}, \dots, e^{a_n t}) = h'_0(t) + \sum_{j=1}^s (h'_j(t) + (a_1 b_{j1} + \dots + a_n b_{jn}) h_j(t)) y_1^{b_{j1}} \cdots y_n^{b_{jn}}$$

Consider the degree and $h_0(t)$ to be non-trivial, it is evident to see that $f_i(t, y_1, ..., y_n) \nmid f'_i(t, y_1, ..., y_n)$. Then $f_i(t, y_1, ..., y_n)$ and $f'_i(t, y_1, ..., y_n)$ are co-prime, since $f_i(t, y_1, ..., y_n)$ is irreducible. For the same reason as above, $f_i(t, e^{a_1t}, ..., e^{a_nt})$ and $f'_i(t, e^{a_1t}, ..., e^{a_nt})$ have no common real root. Therefore, $f_i(t, e^{a_1t}, ..., e^{a_nt})$ has no multiple real root.

4) Complexity analysis of PEFIsolation

Here we give a rough complexity analysis of PEFIsolation. Suppose $f(t) = f_0(t) + f_1(t)e^{v_1t} + \dots + f_s(t)e^{v_st}$, L(f) and U(f) are respectively a lower bound and an upper bound on real roots of f(t), $\deg(f) = (d_0, d_1, \dots, d_s)$. PEFIsolation computes all real roots for a PEF chain f(t) = 0, f'(t) = 0, f''(t) = 0, \dots , totally, $d_0 + \dots + d_{s-1} + s + 1$ such PEFs at most, with the corresponding degree. The last element in the chain is a polynomial with degree d_s , so it has at most d_s real roots. Clearly, for each function in the chain, the number of intervals in its real root isolation is at most $d_0 + d_1 + \dots + d_s + s + 1$. In addition, suppose the lower bound on the distances between real roots of S_i and those of S_{i+1} is δ , then the while loop (line 3-13) in Algorithm 2 always terminates after the length of an interval is less than δ . Since the length of every interval is less than or equal to U(f) - L(f), the while loop must terminate in $\log_2 \frac{U(f) - L(f)}{\delta}$ steps. In a summary, the complexity of PEFIsolation is about $\mathcal{O}((\sum_{i=0}^{s} d_i + s + 1)^2 \log_2 \frac{U(f) - L(f)}{\delta})$.

V. PURELY IMAGINARY EIGENVALUES

In this section, we give a decision procedure for the purely imaginary case described in Section II-C.

A. Reformulation of the Problem

Given an SS as (4), in which all the matrices A_1, \ldots, A_m have only purely imaginary eigenvalues, every component of u_1, u_2, \ldots, u_m is a TMF w.r.t. t and $\bigwedge_{i=1}^m \Lambda(A_i) \cap \Gamma(u_i) = \emptyset$, the initial set X and the unsafe set Y, which both are two open semi-algebraic sets, the goal is to determine whether

$$\mathscr{F}(\mathbf{X},\mathbf{Y}) = \exists \mathbf{x} \exists \mathbf{y} \exists t : \mathbf{x} \in \mathbf{X} \land \mathbf{y} \in \mathbf{Y} \land t \ge 0 \land \Phi(\mathbf{x},t) = \mathbf{y},$$
(24)

holds or not.

B. Solution Form

Theorem 10. Given an SS as (4) as described above, for any initial point $\mathbf{x} \in \mathbb{R}^n$, the solution $\Phi(\mathbf{x}, t)$ is of the following form

$$(\Phi(\mathbf{x},t))_i = \sum_{k=1}^{K_i} z_{ik}^c(\mathbf{x}) \cos(\gamma_{ik}t) + z_{ik}^s(\mathbf{x}) \sin(\gamma_{ik}t), \qquad (25)$$

for i = 1, ..., n, where $z_{ik}^c(\mathbf{x}), z_{ik}^s(\mathbf{x}) \in \mathbb{R}[\mathbf{x}]$ and $\gamma_{ik} \in \mathbb{R}$.

Proof. Similar to Theorem 5.

Example 3. Let

$$\begin{bmatrix} -\xi_2 \\ \xi_1 \\ \xi_2 \\ \xi_3 \\ \xi_4 \end{bmatrix} = \begin{bmatrix} -\xi_2 \\ \xi_1 \\ 2\xi_3 + 2\xi_4 - \xi_1^2 \\ -3\xi_3 - 2\xi_4 + \xi_1\xi_2 \end{bmatrix},$$
(26)

an initial state $\mathbf{x} = (x_1, x_2, x_3, x_4) \in \mathbb{R}^4$, then the solution is

$$\begin{split} \xi_1(t,\mathbf{x}) &= x_1\cos(t) - x_2\sin(t),\\ \xi_2(t,\mathbf{x}) &= x_1\sin(t) + x_2\cos(t),\\ \xi_3(t,\mathbf{x}) &= \frac{\sqrt{2}}{2}(x_1^2 + 2x_1x_2 - 2x_2^2 + 2c + 2d)\sin(\sqrt{2}t) \\ &+ (2x_1x_2 + x_2^2 + c)\cos(\sqrt{2}t) \\ &+ \frac{1}{2}(x_1^2 - 4x_1x_2 - x_2^2)\cos(2t) \\ &- (x_1^2 + x_1x_2 - x_2^2)\sin(2t) - \frac{x_1^2 + x_2^2}{2},\\ \xi_4(t,\mathbf{x}) &= \frac{1}{2}(x_1^2 - 2x_1x_2 - 4x_2^2 + 2x_4)\cos(\sqrt{2}t) \\ &- \frac{\sqrt{2}}{2}(x_1^2 + 4x_1x_2 - x_2^2 + 3x_3 + 2x_4)\sin(\sqrt{2}t) \\ &+ \frac{1}{4}(-5x_1^2 + 5x_2^2 + 4x_1x_2)\cos(2t) \\ &- \frac{1}{2}(x_1^2 - x_2^2 + 5x_1x_2)\sin(2t) + \frac{3}{4}(x_1^2 + x_2^2), \end{split}$$

which is a TMF vector.

As X and Y are two open semi-algebraic sets, there exist some polynomial $p_1(\mathbf{x}), \ldots, p_J(\mathbf{x})$ such that

$$\mathbf{X} = \{ \mathbf{x} \in \mathbb{R}^n \mid p_1(\mathbf{x}) > 0, \dots, p_{J_1}(\mathbf{x}) > 0 \}, \mathbf{Y} = \{ \mathbf{x} \in \mathbb{R}^n \mid p_{J_1+1}(\mathbf{x}) > 0, \dots, p_J(\mathbf{x}) > 0 \}.$$

Thus, the problem (24) can be further reduced to

$$\mathscr{F}(\mathbf{X},\mathbf{Y}) = \exists \mathbf{x} \exists \mathbf{y} \exists t : \Omega, \tag{27}$$

where,

$$\Omega = p_1(\mathbf{x}) > 0, \dots, p_{J_1}(\mathbf{x}) \land p_{J_1+1}(\mathbf{y}) > 0, \dots, p_J(\mathbf{y}) > 0$$

$$\land t \ge 0 \land \bigwedge_{i=1}^n \mathbf{y}_i = \sum_{k=1}^{K_i} z_{ik}^c(\mathbf{x}) \cos(\gamma_{ik}t) + z_{ik}^s(\mathbf{x}) \sin(\gamma_{ik}t).$$
(28)

C. Reduction to Decision Problem of Tarksi's Algebra

In this part, we show the problem (27) can be reduced to the decision problem of Tarski's algebra [39]. There have been many tools available for the decision procedure, e.g., [29], [22], [11], [13], [6], [19], [36], all of which are based on *cylindrical algebraic decomposition* (CAD) [10].

From now on, we will focus on how to reduce (28) to Tarski's algebra equivalently.

For (28), let $\Gamma = \{\gamma_{ik} \mid 1 \le k \le K_i, 1 \le i \le n\}$, i.e., the set of all reals appearing in some trigonometric expression in (28), and $\Delta = \{\delta_1, \ldots, \delta_N\}$ be an integer-basis of Γ , i.e., for any $\gamma \in \Gamma$, γ can be written as a linear combination of Δ with integer coefficients.

In addition, obviously, $\cos(\gamma t)$ and $\sin(\gamma t)$ both are polynomials in $\sin(\delta_1 t), \cos(\delta_1 t), \ldots, \sin(\delta_N t), \cos(\delta_N t)$, for $1 \le k \le K_i, 1 \le i \le n$, that is,

$$\cos(\gamma_{ik}t) = f_{ik}^c(\sin(\delta_1 t), \cos(\delta_1 t), \dots, \sin(\delta_N t), \cos(\delta_N t)), \quad (29)$$

$$\sin(\gamma_{ik}t) = f_{ik}^{s}(\sin(\delta_{1}t), \cos(\delta_{1}t), \dots, \sin(\delta_{N}t), \cos(\delta_{N}t)), \quad (30)$$

where f_{ik}^c , f_{ik}^S are polynomials in their arguments. Denote the following formula by Ξ , i.e.,

$$\Xi \stackrel{c}{=} \mathbf{x} \in \mathbf{X} \wedge \mathbf{y} \in \mathbf{Y} \wedge \bigwedge_{j=1}^{N} u_j^2 + v_j^2 = 1 \wedge$$
$$\bigwedge_{i=1}^{n} y_i = \sum_{k=1}^{K_i} \begin{pmatrix} z_{ik}^c(\mathbf{x}) f_{ik}^c(u_1, v_1, \dots, u_N, v_N) \\ + z_{ik}^c(\mathbf{x}) f_{ik}^c(u_1, v_1, \dots, u_N, v_N) \end{pmatrix}.$$

Theorem 11. Suppose X, Y both are open semi-algebraic sets, Γ is defined as above, which is a set of real numbers, Δ is an integerbasis of Γ , f_{ik}^c and f_{ik}^s are defined as (29), (30), and Ω and Ξ are two formulas defined as above, then

$$\exists \mathbf{x} \exists \mathbf{y} \exists t : \mathbf{\Omega} \Leftrightarrow \exists \mathbf{x} \exists \mathbf{y} \exists_{j=1}^{N} u_{j} \exists_{j=1}^{N} v_{j} : \Xi.$$
(31)

Proof. It is obviously that

$$\exists \mathbf{x} \exists \mathbf{y} \exists t : \mathbf{\Omega} \Rightarrow \exists \mathbf{x} \exists \mathbf{y} \exists_{j=1}^{N} u_{j} \exists_{j=1}^{N} v_{j} : \Xi,$$
(32)

since if there exist $\mathbf{x}, \mathbf{y}, t$ satisfying Ω , let $u_j = \sin(\delta_j t), v_j = \cos(\delta_j t)$, then Ξ is satisfied. So, we just need to prove that

$$\exists \mathbf{x} \exists \mathbf{y} \exists_{j=1}^{N} u_{j} \exists_{j=1}^{N} v_{j} : \Xi \Rightarrow \exists \mathbf{x} \exists \mathbf{y} \exists t : \Omega.$$
(33)

Let

$$S = \{ (\sin(\delta_1 t), \cos(\delta_1 t), \dots, \sin(\delta_N t), \cos(\delta_N t)) \mid t \ge 0 \},\$$

$$\overline{S} = \{ (u_1, v_1, \dots, u_N, v_N) \in \mathbb{R}^{2N} \mid \bigwedge_{i=1}^N u_i^2 + v_i^2 = 1 \}.$$

From Theorem 2, it follows that S is dense in \overline{S} . Denote $\mathbf{w} = (u_1, v_1, \dots, u_N, v_N)$. Let $\mathbf{x}', \mathbf{y}', u_1', \dots, u_N', v_1', \dots, v_N'$ satisfy Ξ , i.e.,

.

$$\begin{split} \mathbf{x}' &\in \mathbf{X} \land \mathbf{y}' \in \mathbf{Y} \land \mathbf{w}' \in S \land \\ &\bigwedge_{i=1}^{n} y_{i}' = \sum_{k=1}^{K_{i}} z_{ik}^{c}(\mathbf{x}') f_{ik}^{c}(\mathbf{w}') + z_{ik}^{s}(\mathbf{x}') f_{ik}^{s}(\mathbf{w}'), \end{split}$$

where $\mathbf{w}' = (u'_1, v'_1, \dots, u'_N, v'_N)$. Since Y is an open set, $\mathbf{y}' \in \mathbf{Y}$, there exists an open ball $B_{\varepsilon}(\mathbf{y}') \subset \mathbf{Y}$, where $B_{\varepsilon}(\mathbf{y}')$ is the ball with center \mathbf{y}' and radius $\varepsilon > 0$. Moreover,

$$y_i = \sum_{k=1}^{K_i} z_{ik}^c(\mathbf{x}') f_{ik}^c(\mathbf{w}) + z_{ik}^s(\mathbf{x}') f_{ik}^s(\mathbf{w}),$$

is a continuous function on \mathbf{w} (denote by $\mathbf{y} = \mathbf{y}(\mathbf{w})$). Thus, there must exist an open ball $B_{\sigma}(\mathbf{w}')$ such that $\mathbf{y}(B_{\sigma}(\mathbf{w}')) \subset B_{\varepsilon}(\mathbf{y}') \subset \mathbf{Y}$, where $\sigma > 0$. Besides, as $\mathbf{w}' \in \overline{S}$ and S is dense in \overline{S} , there must exist $\mathbf{w}_0 \in$ $B_{\sigma}(\mathbf{w}')$, i.e., there exists $t_0 > 0$ with $(a_1t_0, \dots, a_Nt_0) \in B_{\sigma}(\mathbf{w}')$ and $\mathbf{y}_0 = \mathbf{y}(\mathbf{w}_0) \in B_{\varepsilon}(\mathbf{y}') \subset \mathbf{Y}$. Hence, $\mathbf{x}', \mathbf{y}_0, t_0$ satisfy Ω . This completes the proof.

From the decidability of Tarski's algebra [39], an immediate result of Theorem 11 is

Theorem 12. The problem described in (27) is decidable.

Example 4. For a given SS as

$$\begin{pmatrix} \dot{\xi_1} \\ \dot{\xi_2} \end{pmatrix} = \begin{pmatrix} 2 & 2 \\ -3 & -2 \end{pmatrix} \begin{pmatrix} \xi_1 \\ \xi_2 \end{pmatrix} + \begin{pmatrix} \cos(t) \\ \sin(t) \end{pmatrix}$$

and an initial point $\xi(0) = (x_1, x_2)$, its solution is

$$\begin{split} \Phi((x_1, x_2), t) &= \\ \begin{pmatrix} (x_1 + 2)\alpha_1 + \sqrt{2}(x_1 + x_2)\beta_1 - 2\alpha_2 - \beta_2 \\ (x_2 - 2)\alpha_1 - \sqrt{2}(\frac{3}{2}x_1 + x_2 + 1)\beta_1 + 2\alpha_2 + 2\beta_2 \end{pmatrix}, \end{split}$$

where $\alpha_1 = \cos(\sqrt{2}t)$, $\beta_1 = \sin(\sqrt{2}t)$, $\alpha_2 = \cos(t)$, $\beta_2 = \sin(t)$. Given an initial set X and unsafe set Y defined as

$$\begin{aligned} \mathbf{X} &= \{ (x_1, x_2) \mid x_1^2 + x_2^2 < 1 \}, \\ \mathbf{Y} &= \{ (y_1, y_2) \mid y_1 + y_2 > 4 \}, \end{aligned}$$

we want to check whether this system is safe or not. By Theorem 11, we just need to check whether

$$\mathscr{F} := x_1^2 + x_2^2 < 1 \land \alpha_1^2 + \beta_1^2 = 1 \land \alpha_2^2 + \beta_2^2 = 1$$
$$\land (x_1 + x_2)\alpha_1 - \sqrt{2}(\frac{1}{2}x_1 + 1)\beta_1 + \beta_2 > 4,$$

is satisfiable or not. It is easy to prove that there does not exist any $x_1, x_2, \alpha_1, \alpha_2, \beta_1, \beta_2 \in \mathbb{R}$ such that the above formula holds. Thus, the system is safe.

On the other hand, if the unsafe set Y is replaced by

$$\mathbf{Y}' = \{(y_1, y_2) \mid y_1 + y_2 > 3\},\$$

then

$$\mathscr{F}' = x_1^2 + x_2^2 < 1 \land \alpha_1^2 + \beta_1^2 = 1 \land \alpha_2^2 + \beta_2^2 = 1$$

$$\land (x_1 + x_2)\alpha_1 - \sqrt{2}(\frac{1}{2}x_1 + 1)\beta_1 + \beta_2 > 3.$$

Let $x_1 = 0.99$, $x_2 = 0$, $\alpha_1 = \frac{\sqrt{5}}{5}$, $\beta_1 = -\frac{2\sqrt{5}}{5}$, $\alpha_2 = 0$, $\beta_2 = 1$, then $(x_1 + x_2)\alpha_1 - \sqrt{2}(\frac{1}{2}x_1 + 1)\beta_1 + \beta_2 \approx 3.334 > 3$. Thus, the system becomes unsafe.

Example 5. Continue to consider the Example 3. Let the initial set X and unsafe set Y defined as following,

$$X = \{ (x_1, x_2, x_3, x_4) \mid -1 < x_1 < 1 \land x_2 = 0 \land x_3^2 + x_4^2 < 1 \},\$$

$$Y = \{ (y_1, y_2, y_3, y_4) \mid y_3 + y_4 > 4 \},\$$

we want to check whether this system is safe or not. In order to use Theorem 11, we first introduce some new variables as,

$$\alpha_1 = \sin(t), \beta_1 = \cos(t), \alpha_2 = \sin(\sqrt{2}t), q = \cos(\sqrt{2}t)$$

Then the solution of (26) is

$$\begin{split} \xi_1(t,\mathbf{x}) &= x_1\beta_1 - x_2\alpha_1, \\ \xi_2(t,\mathbf{x}) &= x_1\alpha_1 + x_2\beta_1, \\ \xi_3(t,\mathbf{x}) &= \frac{\sqrt{2}}{2}(x_1^2 + 2x_1x_2 - 2x_2^2 + 2c + 2d)\alpha_2 \\ &\quad + (2x_1x_2 + x_2^2 + c)\beta_2 \\ &\quad + \frac{1}{2}(x_1^2 - 4x_1x_2 - x_2^2)(\beta_1^2 - \alpha_1^2) \\ &\quad - 2(x_1^2 + x_1x_2 - x_2^2)\alpha_1\beta_1 - \frac{x_1^2 + x_2^2}{2}, \\ \xi_4(t,\mathbf{x}) &= \frac{1}{2}(x_1^2 - 2x_1x_2 - 4x_2^2 + 2x_4)\beta_2 \\ &\quad - \frac{\sqrt{2}}{2}(x_1^2 + 4x_1x_2 - x_2^2 + 3x_3 + 2x_4)\alpha_2 \\ &\quad + \frac{1}{4}(-5x_1^2 + 5x_2^2 + 4x_1x_2)(\beta_1^2 - \alpha_1^2) \\ &\quad - (x_1^2 - x_2^2 + 5x_1x_2)\alpha_1\beta_1 + \frac{3}{4}(x_1^2 + x_2^2). \end{split}$$

By Theorem 11, we just need to check whether

$$\mathscr{F} := -1 < x_1 < 1 \land x_3^2 + x_4^2 < 1 \land \alpha_1^2 + \beta_1^2 = 1 \land$$
$$\alpha_2^2 + \beta_2^2 = 1 \land \left(\begin{array}{c} \frac{1}{4}(3\alpha_1^2 - 3\beta_1^2 - 4\alpha_1\beta_1 + 1)x_1^2\\ + \frac{1}{2}(x_1^2 + 2x_3 + 2x_4)\beta_2 - \frac{\sqrt{2}}{2}x_3\alpha_2\end{array}\right) > 4,$$

is satisfiable or not. It is easy to prove that there does not exist any $x_1, x_2, x_3, x_4, \alpha_1, \alpha_2, \beta_1, \beta_2 \in \mathbb{R}$ such that the above formula holds. Thus, the system is safe.

Remark 2. Note that the openness of the initial set X and the unsafe set Y plays a very important role in our approach. Otherwise, there may be some point on the boundary of X or Y, which cannot be contained by any ball contained correspondingly in X or Y. But in case either of them is not open, we can resort to the below approach to approximate the reachable set.

VI. ABSTRACTION OF SOLVABLE DYNAMICAL SYSTEMS

In this section, we present an approach to approximate the reachable sets of the general solvable dynamical systems SS (4) by abstracting to the case only with real eigenvalues as discussed in Section IV.

A. Solution Form of the General Case

Given an SS of (4), we will show that its solution is a PETF vector. Namely,

Theorem 13. Given an SS as (4) and an initial point $\mathbf{x} \in \mathbb{R}^n$, then its solution $\Phi(\mathbf{x},t)$ can be represented by the following form

$$(\Phi(\mathbf{x},t))_i = \sum_{k=1}^{K_i} e^{\alpha_{ik}t} (z_{ik}^c(\mathbf{x})\cos(\gamma_{ik}t) + z_{ik}^s(\mathbf{x})\sin(\gamma_{ik}t)), \qquad (34)$$

for i = 1, ..., n, where $z_{ik}^c(\mathbf{x}), z_{ik}^s(\mathbf{x}) \in \mathbb{R}[\mathbf{x}]$ and $\alpha_{ik}, \gamma_{ik} \in \mathbb{R}$.

Proof. Similar to Theorem 5.

B. Approximation of Reachable Sets by Abstraction

Using the solution form above, the reachability of *Y* from *X*, i.e. the safety problem, can be formally described as $\exists \mathbf{x} \exists \mathbf{y} \exists t : \Omega$, where the quantifier free part Ω is defined by

$$\begin{split} \Omega &\doteq \mathbf{x} \in \mathbf{X} \land \mathbf{y} \in \mathbf{Y} \land t \geq 0 \land \\ & \bigwedge_{i=1}^{n} y_{i} = \sum_{k=1}^{K_{i}} e^{\alpha_{ik}t} (z_{ik}^{c}(\mathbf{x})\cos(\gamma_{ik}t) + z_{ik}^{s}(\mathbf{x})\sin(\gamma_{ik}t)). \end{split}$$

The reachability problem of this form is generally undecidable due to the trigonometric functions in the formula. However, if there are no such functions it becomes decidable, and a decision procedure has been proposed in [16]. This fact hints us to eliminate the trigonometric functions by overapproximation of the reachable set, which is analogous to the technique used in Section V. Let

$$\begin{split} \Xi &\doteq \mathbf{x} \in \mathbf{X} \wedge \mathbf{y} \in \mathbf{Y} \wedge t \ge 0 \wedge \bigwedge_{j=1}^{N} u_j^2 + v_j^2 = 1 \wedge \\ & \bigwedge_{i=1}^{n} y_i = \sum_{k=1}^{K_i} e^{\alpha_{ik}t} \left(\begin{array}{c} z_{ik}^c(\mathbf{x}) f_{ik}^c(u_1, v_1, \dots, u_N, v_N) \\ + z_{ik}^s(\mathbf{x}) f_{ik}^s(u_1, v_1, \dots, u_N, v_N) \end{array} \right). \end{split}$$

Then it follows immediately that

Theorem 14. $\exists \mathbf{x} \exists \mathbf{y} \exists t : \Omega \Rightarrow \exists \mathbf{x} \exists \mathbf{y} \exists t \exists_{j=1}^N u_j \exists_{j=1}^N v_j : \Xi.$

Hence, we can conclude, by Theorem 14, the system to be verified is safe, i.e., *Y* is not reachable from *X*, as long as we can prove $\exists \mathbf{x} \exists \mathbf{y} \exists t \exists_{i=1}^{N} u_{j} \exists_{i=1}^{N} v_{j} : \Xi$ does not hold.

VII. IMPLEMENTATION AND EXPERIMENTAL RESULTS

We have discussed the reachability problem for four cases: nilpotent, real, purely imaginary and general case. For the nilpotent and purely imaginary case, the reachability problem can be reduced to a quantifier elimination problem in Tarski's algebra by introducing some new variables. Therefore, it is easy to obtain the decidability since the quantifier elimination of Tarski's algebra is decidable. We implement a tool to decide the reachability problem for the real case. And then using such tool to obtain an incomplete method for the general case.

A. Part 1: Only With Real Eigenvalues

We have implemented the proposed approach for the case only with real eigenvalues described in section IV in Mathematica as a prototype, called LinR 2 , which takes an SS reachability problem as input, and gives either False if the problem is not satisfiable, or True otherwise associated with some valid sample points.

Remark 3. When we implement the above algorithms, some optimizing strategies are adopted for improving efficiency. For example, if the input function can be factorized, then we isolate the real roots of each factor rather than the input function itself, and then refine the resulted intervals if necessary. We omit the implementation details here.

In the following, we report some experimental results with LinR.

Example 6. Consider the following linear dynamical system

$$\dot{\xi} = \begin{bmatrix} \sqrt{2} & & \\ & -\sqrt{2} & \\ & & -1 \end{bmatrix} \xi + \begin{bmatrix} 1-t \\ te^t \\ e^{-t} \end{bmatrix}.$$

Let

1

$$\begin{aligned} \mathbf{X} &= \{ (x_1, x_2, x_3)^T \mid 1 - x_1^2 - x_2^2 - x_3^2 > 0 \}, \\ \mathbf{Y} &= \{ (y_1, y_2, y_3)^T \mid y_1 + y_2 + y_3 + 2 < 0 \}. \end{aligned}$$

The safety property to be verified is to check if some state in Y is reachable from X.

Obviously, $X \cap Y = \emptyset$, and

$$\xi(t) = \begin{bmatrix} x_1 e^{\sqrt{2}t} + \frac{\sqrt{2}t - \sqrt{2} + 1}{2} + \frac{\sqrt{2} - 1}{2} e^{\sqrt{2}t} \\ x_2 e^{-\sqrt{2}t} + \frac{(1 + \sqrt{2})t - 1}{3 + 2\sqrt{2}} e^t + \frac{e^{-\sqrt{2}t}}{3 + 2\sqrt{2}} \\ x_3 e^{-t} + t e^{-t} \end{bmatrix}$$

is the solution of the LDS. Thus, the reachability problem becomes

$$\mathcal{F} = \exists x_1 \exists x_2 \exists x_3 \exists t. \ \Phi(x_1, x_2, x_3, t);$$

$$\Phi(x_1, x_2, x_3, t) = 1 - x_1^2 - x_2^2 - x_3^2 > 0 \land t > 0$$

$$\land x_1 e^{\sqrt{2}t} + x_2 e^{-\sqrt{2}t} + x_3 e^{-t} + h(t) < 0.$$

where $h(t) = \frac{e^{-\sqrt{2}t}}{3+2\sqrt{2}} + te^{-t} + \frac{\sqrt{2}t-\sqrt{2}+5}{2} + \frac{(1+\sqrt{2})t-1}{3+2\sqrt{2}}e^t + \frac{\sqrt{2}-1}{2}e^{\sqrt{2}t}$. Then, using *Brown's projection operator* [19] to eliminate x_1, x_2, x_3

successively (Step 3 in Section IV-C), we have

$$\begin{aligned} q_3(x_1, x_2, x_3, t) &= (x_1^2 + x_2^2 + x_3^2 - 1)(ax_1 + bx_2 + cx_3 + h) \\ q_2(x_2, x_3, t) &= a(x_2^2 + x_3^2 - 1) \\ (-a^2 + a^2 x_2^2 + a^2 x_3^2 + b^2 x_2^2 + 2bcx_2 x_3 + 2bhx_2 + c^2 x_3^2 + 2chx_3 + h^2), \\ q_1(x_3, t) &= a(x_3 - 1)(x_3 + 1)(a^2 + b^2)(2chx_3 + h^2 - b^2 + b^2 x_3^2 + c^2 x_3^2) \\ (-a^2 + a^2 x_3^2 + 2chx_3 + h^2 - b^2 + b^2 x_3^2 + c^2 x_3^2), \end{aligned}$$

$$\begin{aligned} q_0(t) &= ab(c - h)(c + h)(b^2 + c^2 - h^2)(a^2 + b^2 + c^2) \\ (a^2 + b^2)(b^2 + c^2)(a^2 + b^2 + c^2 - h^2). \end{aligned}$$

where $a = e^{\sqrt{2}t}$, $b = e^{-\sqrt{2}t}$ and $c = e^{-t}$.

Isolate all real roots of $q_0(t) = 0$ in $(0, +\infty)$ (as we only care t > 0) (Step 4 in Section 4), and obtain $\mathscr{L}(q_0) = \{(1.08, 1.29)\}.$

Lift the real root isolation in the order t, x_3, x_2, x_1 successively using the openCAD lifting procedure (Step 5 in Section 4), finally, we obtain 48 sample points, in which the sample point $\{-0.835, -0.212, 0.184, 2.\}$ satisfies Φ , which implies that the safety property is not satisfied with the counter example starting from $(-0.835, -0.212, 0.184) \in X$, and ending at time t = 2.

²Both the tool and the case studies in this section can be found at http: //lcs.ios.ac.cn/~chenms/tools/LinR.tar.bz2

Example 7 (Adapted from [1]). Consider a vessel of water containing a radioactive isotope, to be used as a tracer for the food chain, which consists of aquatic plankton varieties phytoplankton A and zooplankton B. Let $\xi_1(t)$ be the isotope concentration in the water, $\xi_2(t)$ the isotope concentration in A and $\xi_3(t)$ the isotope concentration in B. The dynamics of the vessel is modelled as $\dot{\xi} = A\xi$, where $A = \begin{bmatrix} -3 & 6 & 5\\ 2 & -12 & 0\\ 1 & 6 & -5 \end{bmatrix}$. The initial radioactive isotope concentrations $\xi_1(0) = x_1 > 0, \xi_2(0) = 0, \xi_3(0) = 0$.

The safety property of our concern is whether $\forall t > 0 \ \xi_1(t) \geq$ $\xi_2(t) + \xi_3(t)$. To this end, we consider a more general problem: For which $n_1, n_2 \in \mathbb{N}$ s.t. $\mathscr{F}(n_1, n_2) = \exists x_1 > 0 \quad \exists t > 0 \quad \xi_1(t) < 0$ $n_1\xi_2(t) + n_2\xi_3(t)$ holds.

It is easy to see that the matrix A is diagonalizable with eigenvalues $0, -10 + \sqrt{6}, -10 - \sqrt{6}$. When $(n_1, n_2) = (1, 1)$, using the method in Section IV-C, we obtain two sample points for (x_1,t) , i.e., (-0.1,1), (0.1,1). But none of them satisfies $\mathscr{F}(1,1)$, which simply implies the safety property holds. When $(n_1, n_2) = (2, 2)$, similarly, we obtain four sample points for (x_1,t) , i.e., (-0.1,0), (0.1,0), (-0.1, 1), (0.1, 1), in which (0.1, 1) satisfies $\mathscr{F}(2, 2)$. It can be proved that $\xi_i(t) \ge 0$ for any t > 0 and i = 1, 2, 3. So, it is clear that, if $\mathscr{F}(n_1, n_2)$ holds, $\mathscr{F}(m_1, m_2)$ holds for $m_1 \ge n_1$ and $m_2 \ge n_2$. Then, by checking some pairs of $(n_1, n_2) \in \mathbb{N} \times \mathbb{N}$ in a similar way as above, we conclude that all pairs $(n_1, n_2) \in \mathbb{N} \times \mathbb{N}$ satisfy $\mathscr{F}(n_1, n_2)$, except for the pairs $\{(0,0), (0,1), (0,2), (1,0), (1,1), (1,2), (2,0), (2,1), (2,1),$ $(3,0),(3,1),(4,0),(5,0)\}.$

Example 8 ([Adapted from [1]). Consider a typical home with attic, basement and insulated main floor. Let $x_3(t), x_2(t), x_1(t)$ be the temperature in the attic, main living area and basement respectively, and t is the time in hours. Assume it is winter time, the outside temperature is nearly $35^{\circ}F$, and the basement earth temperature is nearly 45°F. Suppose a small electric heater is turned on, and it provides a $20^{\circ}F$ rise per hour. We want to verify that the temperature in main living area will never reach too high (maybe $70^{\circ}F$). Analyze the changing temperatures in the three levels using Newton's cooling law and given the value of the cooling constants, we obtain the model as follows:

$$\begin{split} \dot{x_1} &= \frac{1}{2}(45 - x_1) + \frac{1}{2}(x_2 - x_1), \\ \dot{x_2} &= \frac{1}{2}(x_1 - x_2) + \frac{1}{4}(35 - x_2) + \frac{1}{4}(x_3 - x_2) + 20, \\ \dot{x_3} &= \frac{1}{4}(x_2 - x_3) + \frac{3}{4}(35 - x_3), \end{split}$$

with the initial set $\mathbf{X} = \{(x_1, x_2, x_3)^T \mid 1 - (x_1 - 45)^2 - (x_2 - 35)^2 - (x_2 - 35)^$ $(x_3 - 35)^2 > 0$ and the unsafe set $Y = \{(y_1, y_2, y_3)^T \mid y_2 - 70 > 0\}.$ The safety property we are concerning is to check if some state in Y is reachable from X, which holds by using LinR.

Example 9. Consider a non-linear SS as follows,

$$\begin{split} \dot{\xi_1} &= -\xi_1 + 2\xi_2, \\ \dot{\xi_2} &= \xi_1 - \xi_2, \\ \dot{\xi_3} &= -\xi_3 + \xi_1\xi_2 \end{split}$$

with the initial set $\mathbf{X} = \{(x_1, x_2, x_3)^T \mid -x_1 + x_2 - x_3 + 2 < 0\}$ and the unsafe set $Y = \{(y_1, y_2, y_3)^T \mid -y_1 + y_2 - y_3 - 2 > 0\}$. For an initial point $x = (x_1, x_2, x_3)$, the solution is,

$$\begin{split} \xi_1(t,x) &= (\frac{1}{2}a + \frac{\sqrt{2}}{2}b)e^{(\sqrt{2}-1)t} - (-\frac{1}{2}a + \frac{\sqrt{2}}{2}b)e^{-(\sqrt{2}+1)t} \\ \xi_2(t,x) &= (\frac{\sqrt{2}}{4}a + \frac{1}{2}b)e^{(\sqrt{2}-1)t} + (-\frac{\sqrt{2}}{4}a + \frac{1}{2}b)e^{-(\sqrt{2}+1)t} \\ \xi_3(t,x) &= e^{-t}(\frac{e^{(2\sqrt{2}-1)t}}{2\sqrt{2}-1}(\frac{\sqrt{2}}{8}x_1^2 + \frac{\sqrt{2}}{4}x_2^2 + \frac{1}{2}x_1x_2) \\ &+ \frac{e^{-(2\sqrt{2}+1)t}}{2\sqrt{2}+1}(\frac{\sqrt{2}}{8}x_1^2 + \frac{\sqrt{2}}{4}x_2^2 - \frac{1}{2}x_1x_2) \\ &+ c - \frac{1}{8}\frac{\sqrt{2}x_1^2}{2\sqrt{2}-1} - \frac{1}{4}\frac{\sqrt{2}x_2^2}{2\sqrt{2}-1} - \frac{1}{2}\frac{x_1x_2}{2\sqrt{2}-1} \\ &+ \frac{1}{8}\frac{\sqrt{2}x_1^2}{-2\sqrt{2}-1} + \frac{1}{4}\frac{\sqrt{2}x_2^2}{-2\sqrt{2}-1} - \frac{1}{2}\frac{x_1x_2}{-2\sqrt{2}-1}) \end{split}$$

The safety property we are concerning is to check if some state in Y is reachable from X, i.e., check whether the following formula is true of not,

$$\exists x_1 \exists x_2 \exists x_3 \exists t : -x_1 + x_2 - x_3 + 2 < 0 \land t \ge 0 \\ \land -\xi_1 + \xi_2 - \xi_3 - 2 > 0.$$

Using our tool, a point $(x_1, x_2, x_3, t) = (-36.1203, 20.7631, 59.1, 1)$ can be found that satisfy the above formula, which means that the system will reach Y from the initial point $(-36.1203, 20.7631, 59.1) \in$ X at time t = 1. Thus, it is unsafe.

The above four examples are verified by LinR. Both the time and memory costs on a 64-bit Linux computer with a 2.93GHz Intel Core-i7 processor and 4GB of RAM are shown in Table I. Besides, we have also compared on the same platform with the performance of Strzeboński's approach (i.e., CT1D) [38], as well as verification tools dReach [25], HSolver [33], and Flow* [7] on these examples. Note that, both dReach and Flow* cannot handle unbounded model checking, and even for BMC, they are less efficient than our tool in many cases (see Example 6, Example 7 and Example 9)³. In particular, Flow* accepts only rectangular initial set, i.e. each variable needs to be specified within a closed interval and polynomial constraints are not allowed, and thus we tried different cube to approximate the spherical initial set in Example 6, while none of them can derive a desired result ("unsafe"). As for HSolver, due to the rejection of "sqrt", we simplify the original model by replacing all the irrational numbers with their approximate decimals, however, 2 of the 3 examples still can not be answered by HSolver in reasonable time and memory.

Remark 4. In the above examples, all constraints are open sets. Actually, more general initial and unsafe sets, i.e., when either Pre(X)or Post(X) is not open semi-algebraic, can be coped with in our approach also, as we have implemented CAD in the algorithms. For the Example 3.4 in [27], where A is diagonalizable with rational eigenvalues and Pre(X) and Post(X) are both closed sets, it takes 57 milliseconds using Lafferriere et al.'s approach based on quantifier elimination by QEPCAD [11]. In contrast, LinR takes 39 milliseconds, and CT1D takes 33 milliseconds. In brief, our approach shares nearly same complexity as Strzeboński's in general case, but is still better than other approaches, see Table II (QEPCAD stands for Lafferriere et al.'s approach).

Remark 5. It is worth clarifying the aim of the comparison done in this section, though we recognize that the comparisons with dReach, Flow* and HSolver are not essentially fair in general, due to distinction of their scopes. A more reasonable way of doing the comparison might be with several state-of-the-art tools for quantifier elimination, e.g. REDLOG [14], QEPCAD, and SyNRAC [24]. However, these implementations are not capable of dealing with the examples listed in Table I, as we are considering more general classes of systems featuring decidability results. For instance, SyNRAC performs quantifier elimination only over polynomial formulas, yet not available for constraints involving transcendental functions. Whilst, CT1D, a generalized CAD implementation of Mathematica's Reduce command, is theoretically competent in solving those examples, and thus is listed as one of the candidates in Table I.

Aiming at an extensive evaluation of our algorithms, especially for the efficiency, we resort to the verification community by performing comparisons with tools therein for reachability computation. Whereas unfortunately, neither dReach, Flow*, HSolver nor SpaceEx [15] is fully compatible with our examples, and therefore some simplifications or approximations over the examples are conducted before triggering those tools. For instance, we feed dReach and Flow* with a time bound respectively for each example, as they cannot handle unbounded verification; we replace the unbounded initial set with a small compact one in Example 7 and 9 when evaluating HSolver, dReach and Flow*, due to their intractability of unbounded initial set; while a rectangular approximation of the initial set is always needed for Flow* if the variables are not originally specified within closed intervals.

Particularly, for systems considered in this paper, if no simplification or approximation techniques are involved, one could get an immediate overview of the advantages of our approach through Table III.

B. Part 2: Abstraction of Solvable Dynamical Systems

To demonstrate the effectiveness of our technique which uses abstraction for general solvable dynamical systems with complex eigenvalues, we have extended our tool called LinR [16] in Mathematica, which has been demonstrated more efficient than existing approaches based on approximation and numeric computation in general, e.g., HSolver, dReach, FLOW*, etc. For systems with real or purely imaginary eigenvalues, the tool produces an exact result in finite time declaring the system "SAFE" or "UNSAFE"; while for systems with complex eigenvalues where overapproximation is used, the algorithm is guaranteed to terminate in a finite number of steps, either by finding a real counterexample (sample point) in the concrete system and declaring the system "UNSAFE", or by claiming the system "SAFE" when the abstracted system is safe, i.e. no counterexample is detected, or returning an "UNKNOWN" answer when the abstracted system is unsafe but the concrete system is safe, where only spurious counterexamples can be derived. In what follows, we illustrate our approach by several real-world examples.

1) Pond Pollution

Consider three ponds connected by streams, where the first pond has an external pollution source that spreads via the connecting streams to the other two ponds. Denote $x_1(t), x_2(t), x_3(t)$ as the amount (lbs) of pollutant in ponds 1, 2, 3 respectively, and t as the time in minutes. Assume that the pollutant is well-mixed in each pond, and we plan to verify that the amount of pollutant in pond 2 stays higher than that in pond 3 with an offset, 6 lbs for instance. By using a compartment analysis and instantiating the parameters⁴, we

³Here, we set the time bounds 2s, 2s, 5s and 2s resp. for examples 6, 7, 8 and 9 when using dReach and Flow*.

⁴For more details, please refer to http://www.math.utah.edu/~gustafso/ s2013/2250/systemsExamplesTheory2008.pdf

LDS	Time (sec)				Memory (kb)					
	LinR	CT1D	dReach	HSolver	Flow*	LinR	CT1D	dReach	HSolver	Flow*
Example 6	1.35	×	37.36	_	-	112	×	3812	-	-
Example 7	0.03	0.20	0.71	_	_	131	2018	3816	-	-
Example 8	1.68	×	0.05	0.72	16.50	166	×	3812	1076932	113492
Example 9	17.56	×	22.48	-	-	580	×	3820	-	-
\times : the verification fails by non-termination within reasonable amount of time (10 hours)										

× . the vertication rans by non-termination within reasonable amount of time (10 hou

- : the verification fails because of giving an answer as "safety unknown"

EVALUATION RESULTS OF DIFFERENT METHODS

LinR	CT1D	QEPCAD	dReach	HSolver	Flow*
39	33	57	110	_	-

 TABLE II

 Time consumption (in milliseconds) on Example 3.4 from [27]

Features	LinR	HSolver	dReach	Flow*	SpaceEx	
unbounded time verification			-	-	$^{\#}$	
unbounded initial set		_	_	-	-	
non-linear semi-algebraic initial set		\checkmark	\checkmark	-	_	
non-linear solvable systems	\checkmark			\checkmark	-	

 $\sqrt{\pi}$: based on existence of fixed-points of the reachable states

 TABLE III

 Features supported by different tools

obtain the specialized dynamics as

$$\begin{aligned} \dot{x}_1(t) &= 0.001x_3(t) - 0.001x_1(t) + 0.01, \\ \dot{x}_2(t) &= 0.001x_1(t) - 0.001x_2(t), \\ \dot{x}_3(t) &= 0.001x_2(t) - 0.001x_3(t), \end{aligned}$$

with the initial set $X = \{(x_1, x_2, x_3)^T | (x_1 - 1)^2 + (x_2 - 1)^2 + (x_3 - 1)^2 < 1\}$ and the unsafe set $Y = \{(y_1, y_2, y_3)^T | y_2 - y_3 + 6 < 0\}$. The safety property we are concerning is to check if some state in Y is reachable from X. Since $X \cap Y = \emptyset$, we need further reduce the reachability problem to a quantifier elimination problem.

Observe that the system matrix is diagonalizable with three complex eigenvalues 0, $(-3 - i\sqrt{3})/2000$, and $(-3 + i\sqrt{3})/2000$. By using the solution of this system w.r.t. an initial state $(x_1, x_2, x_3)^T \in X$, the reachability problem thus becomes

$$\mathscr{F} \triangleq \exists x_1 \exists x_2 \exists x_3 \exists t : (x_1 - 1)^2 + (x_2 - 1)^2 + (x_3 - 1)^2 - 1 < 0$$

$$\land a + b \cos\left(\frac{\sqrt{3}t}{2000}\right) + c \sin\left(\frac{\sqrt{3}t}{2000}\right) < 0 \land t > 0,$$

where the second constraint corresponds to the unsafe set Y, with $a = 28e^{3t/2000}$, $b = 3x_2 - 3x_3 - 10$, and $c = \sqrt{3}(2x_1 - x_2 - x_3 - 10)$.

To further reduce the above problem to Tarski's algebra with exponentiations, we abstract the second constraint by eliminating trigonometric functions with overapproximation, *i.e.*

$$a + bu + cv < 0 \land u^2 + v^2 = 1.$$
 (35)

As a quantifier elimination procedure, we can eliminate u and v in (35) by using the Cauchy-Schwarz inequality and thus get

$$a^2 - b^2 - c^2 < 0. (36)$$

The reduced reachability problem is then successfully solved in *LinR* due to its kernel that implements CAD. The original system is verified to be safe inasmuch as no counterexamples of the abstracted system is derived, namely the overapproximation of the original system is safe. In a more intuitive way, Fig. 1 depicts the overapproximation (the tube) of one single trajectory (the curve) starting from $(1,1,1)^T$



Fig. 1. Overapproximation of the trajectory starting from $(1,1,1)^T$

initially. Note that the approximation tends to be tighter as the system evolves along with time, which is essentially on account of the intrinsic convergence of the original system. In other words, the system matrix has three eigenvalues whose real parts are all non-positive. This implies the stability property and thus makes our approach more competitive for checking properties in terms of a long span of time.

2) PID Controller

Consider a proportional-integral-derivative (PID) controller (taken from [31]) which is used to control a simple mass, spring, and damper problem. The modelling equation of the mass, spring, and damper system (plant) is

$$M\ddot{x} + b\dot{x} + kx = F$$

where M = 1kg, b = 10Ns/m, k = 20N/m are given parameters of the plant, and *F* is the controllable force. Suppose the goal is to control the plant to reach a steady state where x = 1 with some requirements on the overshoot and rise time. Let r(t) denote the desired trajectory

for reaching the steady state x = 1, which follows as a step function: r(t) = 0 for t < 0 and r(t) = 1 for t > 0.

Given a PID controller, the model describing the composed plant and controller is

$$M\ddot{x} + b\dot{x} + kx = K_d(r - x) + K_p(r - x) + K_i \int (r - x)^{-1} dx dx$$

where K_d , K_p and K_i are parameters indicating gains of the derivative, proportional and integral respectively, while r - x is the error in tracking the desired trajectory r.

We consider the case of using a PI controller, *i.e.* $K_d = 0$, and choose $K_p = 350$ and $K_i = 300$. We will prove the following property of the system using our approach:

$$\mathbf{G}(t > 0.5 \Rightarrow x \ge 0.9 \land x \le 1.1). \tag{37}$$

Note that this case has been studied in [31] but unfortunately it cannot be proved by the method proposed there.

Let $\mathbf{x} = [\int x, x, \dot{x}, t]^{\mathrm{T}}$, then $\dot{\mathbf{x}} = A\mathbf{x} + \mathbf{u}$, where

$$A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ -300 & -370 & -10 & 300 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

and $\mathbf{u} = [0,0,350,1]^{\mathrm{T}}$. The initial value is $\mathbf{x}(0) = [0,0,0,0]$ and unsafe set is $Y = \{\mathbf{x} \mid t > 0.5 \land (x < 0.9 \lor x > 1.1)\}$. Now the problem has been written in the form of reachability of an LDS. The eigenvalues of *A* are $0, \lambda_0, \lambda_1$, and λ_2 , where λ_i (i = 0, 1, 2) are roots of the characteristic equation $f(\lambda) = \lambda^3 + 10\lambda^2 + 370\lambda + 300$. Solving the LDS we get

$$x = 1 + c_0 \lambda_0 e^{\lambda_0 t} + c_1 \lambda_1 e^{\lambda_1 t} + c_2 \lambda_2 e^{\lambda_2 t},$$

where

$$\begin{bmatrix} c_0 \\ c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ \lambda_0 & \lambda_1 & \lambda_2 \\ \lambda_0^2 & \lambda_1^2 & \lambda_2^2 \end{bmatrix}^{-1} \begin{bmatrix} 1/15 \\ -1 \\ 0 \end{bmatrix}$$

Observe that $f(\lambda)$ has only one real root, denoted by λ_0 , and by λ_1 and λ_2 the other two conjugate complex roots. Let $\lambda_{1,2} = \alpha \pm \beta \mathbf{i}$, then the solution can be rewritten as

$$x = 1 + c_0 \lambda_0 e^{\lambda_0 t} + 2e^{\alpha t} (\operatorname{Re}(c_1 \lambda_1) \cos(\beta t) - \operatorname{Im}(c_1 \lambda_1) \sin(\beta t)).$$

Now by abstraction, we put $u = \cos(\beta t)$, $v = \cos(\beta t)$ and require that $u^2 + v^2 = 1$. Then the reachability of *Y* becomes

$$\exists u \exists v \exists t : u^2 + v^2 = 1 \land t > 0.5 \land (\phi(u, v, t) < -0.1 \lor \phi(u, v, t) > 0.1),$$
(38)

where $\phi(u, v, t) = c_0 \lambda_0 e^{\lambda_0 t} + 2(\operatorname{Re}(c_1 \lambda_1) u - \operatorname{Im}(c_1 \lambda_1) v) e^{\alpha t}$. Then using the method proposed in [16], we prove that (i) $\phi(u, v, t) > 0.1$ is invalid, and thus $x \le 1.1$ in Eq. (37) is verified; and (ii) the interval (0.5, T] covers all *t* that make $\phi(u, v, t) < -0.1$ satisfiable in Eq. (38). Here *T* is the unique root of $|c_0 \lambda_0| e^{\lambda_0 t} + 2|c_1 \lambda_1| e^{\alpha t} - 0.1$, which can be approximated by real root isolation with arbitrary precision. We adopt 0.6 as an overapproximation of *T* here (see Fig. 2).

Using our method it has been shown that *Y* can only be reached when *t* is in (0.5, 0.6]. Moreover, it can be checked by bounded model checking or simulation based verification [18], [23] that even for $t \in (0.5, 0.6]$ *Y* can not be reached. Therefore, we have proved the property (37) for the given system.



Fig. 2. Overapproximation (the "broom") of the trajectory of x (the curve) starting from 0. Here the two horizontal dashed lines specify the boundaries of the safe set, while T indicates a point in time, after which the behaviour of the overapproximated system stays within the safe region.

VIII. CONCLUSION

In this paper, we extended our previous approaches on reachability analysis for linear vector fields given in [16], [17] to solvable nonlinear vector fields. To this end, we first identified three families of solvable non-linear vector fields, i.e., the cases when the matrices in (4) are respectively *nilpotent*, *only with real eigenvalues* and *only with pure imaginary eigenvalues*, and proved their reachability problems are decidable. In addition, we presented an approach on how to abstract the reachability problem of general solvable dynamical systems (4) to the decision problem of \mathscr{T}_e . A prototypical tool has been implemented, and experimental results indicate our approach is efficient.

As a future work, it could be interesting to investigate whether the reachable set computation of general non-linear vector fields, even non-polynomial vector fields can be abstracted to that of solvable ones, further to that of linear ones, by exploiting our previous work in [28].

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