

Algebraic Tools for Computing Polynomial Loop Invariants

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ABSTRACT

Loop invariants are properties of a program loop that hold before and after each iteration of the loop. They are often employed to verify programs and ensure that algorithms consistently produce correct results during execution. Consequently, the generation of invariants becomes a crucial task for loops. We specifically focus on polynomial loops, where both the loop conditions and assignments within the loop are expressed as polynomials. Although computing polynomial invariants for general loops is undecidable, efficient algorithms have been developed for certain classes of loops. For instance, when all assignments within a while loop involve linear polynomials, the loop becomes solvable. In this work, we study the more general case where the polynomials exhibit arbitrary degrees.

Applying tools from algebraic geometry, we present two algorithms designed to generate all polynomial invariants for a while loop, up to a specified degree. These algorithms differ based on whether the initial values of the loop variables are given or treated as parameters. Furthermore, we introduce various methods to address cases where the algebraic problem exceeds the computational capabilities of our methods. In such instances, we identify alternative approaches to generate specific polynomial invariants.

CCS CONCEPTS

- **Applied computing** → **Invariants; Logic and verification** ;
- **Computing methodologies** → **Symbolic and algebraic manipulation**.

KEYWORDS

Program synthesis, Loop invariants, Polynomial ideals

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1 INTRODUCTION

Loop invariants denote properties that hold both before and after each iteration of a loop within a given program. They play a crucial role in automating the program verification, ensuring that algorithms consistently yield correct results prior to execution. Notably, various recognized methods for safety verification like the Floyd–Hoare inductive assertion technique [11] and the termination verification via standard ranking functions technique [21] rely on loop invariants to verify correctness, ensuring complete automation in the verification process.

In this work, we focus on polynomial loops, wherein expressions within assignments and conditions are polynomials equations in program variables. More precisely, a polynomial loop is of the form:

$$\begin{array}{l} (x_1, x_2, \dots, x_n) = (a_1, a_2, \dots, a_n) \\ \textbf{while } g_1 = \dots = g_k = 0 \textbf{ do} \\ \quad \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \leftarrow \textbf{F} \begin{pmatrix} f_1 \\ f_2 \\ \vdots \\ f_n \end{pmatrix} \\ \textbf{end while} \end{array}$$

where the x_i 's represent program variables with initial values a_i , and g_i 's and f_i 's are polynomials in the program variables. Computing polynomial invariants for loops has been a subject of study over the past two decades, see e.g. [1, 10, 14, 16, 18–20, 25–27]. Computing polynomial invariants for general loops is undecidable [15]. Therefore, particular emphasis has been placed on specific families of loops, especially those in which the assertions are all linear or can be reduced to linear assertions. In the realm of linear invariants, Michael Karr introduced an algorithm pioneering the computation of all linear invariants for loops where each assignment within the loop is a linear function [16]. Subsequent studies, such as [22] and [26], have demonstrated the feasibility of computing all polynomial invariants up to a specified degree for loops featuring linear assignments. Further, the problem of generating *all* polynomial invariants for loops with linear assignments, are studied in [14] and [26].

Another class of loops for which invariants have been successfully computed is the family of solvable loops. These loops are characterized by polynomial assignments that are either inherently linear or can be transformed into linear forms through a change of variables, as elaborated in [9] and [19]. Nonetheless, challenges persist when dealing with loops featuring non-linear or unsolvable assignments, as discussed in [27] and [1].

Before stating our main results, we introduce some terminology from algebraic geometry. We refer to [6], [17] for further details. Let \mathbb{C} denote the field of complex numbers. Let S be a set of polynomials in $\mathbb{C}[x_1, \dots, x_n]$, then the *algebraic variety* $V(S)$ associated to S is the common zero set of all polynomials in S . Here, $V(S) = V(\langle S \rangle)$,

where $\langle S \rangle$ is the ideal generated by S . Conversely, the *defining ideal* of a subset $X \subset \mathbb{C}^n$ is the set of polynomials in $\mathbb{C}[x_1, \dots, x_n]$ that vanish on X . The algebraic variety associated to the ideal $I(X)$ is called the *Zariski closure* of X . Therefore, if X is an algebraic variety, then $V(I(X)) = X$. Moreover, $X_1 \subseteq X_2$ implies that $I(X_2) \subseteq I(X_1)$.

A map $F : \mathbb{C}^n \rightarrow \mathbb{C}^m$ is called a *polynomial map*, if there exist f_1, \dots, f_m in $\mathbb{C}[x_1, \dots, x_n]$, such that $F(x) = (f_1(x), \dots, f_m(x))$ for all $x \in \mathbb{C}^n$. For the sake of simplicity, in what follows we will refer to polynomial maps and their associated polynomials interchangeably.

We now define the main object introduced in this paper.

Definition 1.1. Let $F : \mathbb{C}^n \rightarrow \mathbb{C}^m$ be a polynomial map and $X \subseteq \mathbb{C}^m$ an algebraic variety. The invariant set of (F, X) is defined as:

$$S_{(F,X)} = \{x \in X \mid \forall m \in \mathbb{N}, F^{(m)}(x) \in X\},$$

where $F^{(0)}(x) = x$ and $F^{(m)}(x) = F(F^{(m-1)}(x))$ for any $m > 1$.

Our contributions. In this work, we consider the problem of generating polynomial invariants for loops with polynomial maps of arbitrary degrees. Our contributions are as follows:

- (1) We design an algorithm (Algorithm 1) for computing invariant sets and use it to efficiently decide if a given polynomial is invariant (Proposition 3.4)
- (2) We also design two algorithms for computing the set of polynomial invariants of a loop up to a fixed degree. The first one (Theorem 3.5), when the initial value is not fixed, outputs a linear parametrization which depends polynomially on this value. The second (Algorithm 2), when the initial value is fixed, is much more efficient and computes a basis of this set, seen as a vector space. Experiments with our prototype implementation demonstrate the practical efficiency of our algorithms, solving problems beyond the current state of the art.
- (3) Finally, we apply these algorithms to other problems: we show how to lift some polynomial invariants for the non-fixed initial value case from the fixed one (Proposition 4.1); we consider the case with inequalities in the loop (Proposition 4.3).

Related works. A common approach for generating polynomial invariants entails creating a recurrence relation from a loop, acquiring a closed formula for this recurrence relation, and then computing polynomial invariants by removing the loop counter from the obtained closed formula (as in [26]). Note that it is straightforward to find such recursion formulas from a polynomial invariant. However, the reverse process is only feasible under very strong assumptions, as detailed in [1]. Specifically, one needs to identify linear relations among program variables, which is a challenging task in itself.

In [14], an algorithm is designed to compute the Zariski closure of points generated by affine maps. Another perspective, detailed in [1], categorizes variables into effective and defective groups, where closed formulas can be computed for effective variables but not for defective variables. In [7], the method of converting loops into solvable loops is discussed, though such conversion may not always be feasible for non-zero solvable loops. Similarly, the methodology proposed in [19] is specifically tailored for P-solvable loops.

Structure of the paper. In Section 2, we introduce our approach to computing invariant sets using algebraic tools, Algorithm 1. The process of generating polynomial invariants using these invariant sets is detailed in Section 3, accompanied by illustrative examples. In Section 4, we demonstrate some applications of Algorithm 2 in

various examples from the literature. In Section 5, we present tables summarizing our experiments and implementation details.

2 COMPUTING INVARIANT SETS

In this section, we will first establish an effective description of the invariant set associated with a given algebraic variety and a polynomial map. Subsequently, we will derive an algorithm based on this description to compute such a set.

We begin with a technical lemma to express the preimage of an algebraic variety under a polynomial map.

Lemma 2.1. Given a polynomial map $F : \mathbb{C}^n \rightarrow \mathbb{C}^m$ and an algebraic variety $X \subset \mathbb{C}^m$, the preimage $F^{-1}(X)$ is also an algebraic variety. Moreover, if $X = V(g_1, \dots, g_k)$ and $F = (f_1, \dots, f_m)$, where $g_1, \dots, g_k \in \mathbb{C}[y_1, \dots, y_m]$ and $f_1, \dots, f_m \in \mathbb{C}[x_1, \dots, x_n]$, then

$$F^{-1}(X) = V(g_1(f_1, \dots, f_m), \dots, g_k(f_1, \dots, f_m)) \subset \mathbb{C}^n.$$

PROOF. Let $F = (f_1, \dots, f_m)$ and $X = V(g_1, \dots, g_k)$ as in the statement, by definition

$$\begin{aligned} F^{-1}(X) &= \{x \in \mathbb{C}^n \mid (f_1(x), \dots, f_m(x)) \in V(g_1, \dots, g_k)\} \\ &= \{x \in \mathbb{C}^n \mid \forall 1 \leq i \leq k, g_i(f_1(x), \dots, f_m(x)) = 0\}. \end{aligned}$$

Let $h_i = g_i(f_1, \dots, f_m) \in \mathbb{C}[x_1, \dots, x_n]$, for all $1 \leq i \leq k$, then $F^{-1}(X) = V(h_1, \dots, h_k)$, and $F^{-1}(X)$ is an algebraic variety. \square

Before proving the main result, we need the following lemma.

Lemma 2.2. Let $F : \mathbb{C}^n \rightarrow \mathbb{C}^m$ be a polynomial map and $X \subseteq \mathbb{C}^m$ an algebraic variety. Then, $F(S_{(F,X)})$ is a subset of $S_{(F,X)}$.

PROOF. Let $x \in S_{(F,X)}$. By the definition of the invariant set, $F^{(m)}(x) \in X$ for every m . Thus, $F^{(m)}(F(x)) \in X$, implying that $F(x) \in S_{(F,X)}$ for every $x \in S_{(F,X)}$. Hence, $F(S_{(F,X)}) \subseteq S_{(F,X)}$. \square

We now give an effective method to compute invariant sets, by means of a stopping criterion for the intersection of the iterated preimages. In the following, $(F^{(m)})^{-1}$ will be denoted by $F^{(-m)}$.

Proposition 2.3. Let $F : \mathbb{C}^n \rightarrow \mathbb{C}^m$ be a polynomial map and $X \subseteq \mathbb{C}^m$ an algebraic variety. We define $X_m = \bigcap_{i=0}^m F^{-i}(X)$ for all $m \in \mathbb{N}$. Then, the following statements are true:

- (a) $X_{m+1} \subseteq X_m$ for all m .
- (b) There exists $N \in \mathbb{N}$ such that $X_N = X_m$ for all $m \geq N$.
- (c) If $X_N = X_{N+1}$ for some N , then $X_N = X_m$ for all $m \geq N$.
- (d) The invariant set $S_{(F,X)}$ is equal to X_N .

PROOF. (a) The following is straightforward from the definition:

$$X_{m+1} = X_m \cap F^{-(m+1)}(X) \subseteq X_m.$$

(b) From (a), we have the following descending chain

$$X_0 \supseteq X_1 \supseteq X_2 \supseteq \dots \supseteq X_m \supseteq X_{m+1} \supseteq \dots,$$

which are algebraic varieties by Lemma 2.1. Thus, we have:

$$I(X_0) \subseteq I(X_1) \subseteq I(X_2) \subseteq \dots \subseteq I(X_m) \subseteq I(X_{m+1}) \subseteq \dots$$

Since $\mathbb{C}[x_1, x_2, \dots, x_n]$ is a Noetherian ring, there exists a natural number N such that $I(X_N) = I(X_m)$ for all $m \geq N$. Therefore,

$$X_N = V(I(X_N)) = V(I(X_m)) = X_m \text{ for all } m \geq N.$$

(c) For such an N , we have that

$$X_{N+2} = X \cap F^{-1}(X_{N+1}) = X \cap F^{-1}(X_N) = X_{N+1}.$$

Thus, $X_m = X_{m+1}$ for all $m \geq N$, and so $X_N = X_m$ for all $m \geq N$.

(d) We will first prove that $S_{(F,X)} \subseteq X_m$ for every m , by induction on m . By the invariant set's definition, $S_{(F,X)}$ is a subset of $X = X_0$ which proves the base case $m = 0$. Now let $m > 0$ and assume $S_{(F,X)} \subseteq X_{m-1}$. By Lemma 2.2 and the induction hypothesis,

$$F(S_{(F,X)}) \subseteq S_{(F,X)} \subseteq X_{m-1}.$$

Therefore, $S_{(F,X)}$ is a subset of $F^{-1}(X_{m-1})$. Note that $S_{(F,X)}$ is a subset of X by the definition. Thus, we have

$$S_{(F,X)} \subseteq F^{-1}(X_{m-1}) \cap X = X_m.$$

In particular, when $m = N$, we have that $S_{(F,X)} \subseteq X_N$.

To prove the other inclusion, for every $x \in X_m$, note that $F^m(x)$ is contained in X since $x \in X_m \subseteq F^{-m}(X)$. By (a) and (b), X_N is contained in X_m for every $m \in \mathbb{N}$. Thus, $F^m(x)$ is contained in X for every $x \in X_N$ and every $m \in \mathbb{N}$. Hence, $X_N \subseteq S_{(F,X)}$. \square

Remark 1. By Theorem 2.3(d), the invariant set $S_{(F,X)}$ is an algebraic variety, since by construction each X_i is an algebraic variety. By Theorem 2.3(a), the ideal of X_j is a subset of the ideal of X_i for every $i \geq j$. Hence, although computing the ideal of X_N where $X_N = X_{N+1}$ may be infeasible, leveraging computable X_i 's for $i < N$ provides partial information.

We now present an algorithm for computing the invariant set associated to an algebraic variety and a polynomial map, described by sequences of multivariate polynomials, with *rational* coefficients.

Algorithm 1 InvariantSet

Input: Two sequences \mathbf{g} and $F = (f_1, \dots, f_n)$ in $\mathbb{Q}[x_1, \dots, x_n]$.

Output: Polynomials whose common zero-set is $S_{(F,V(\mathbf{g}))}$.

```

1:  $S \leftarrow \{\mathbf{g}\}$ ;
2:  $\tilde{\mathbf{g}} \leftarrow \text{Compose}(\mathbf{g}, F)$ ;
3: while  $\text{InIdeal}(\tilde{\mathbf{g}}, S) == \text{False}$  do
4:    $S \leftarrow S \cup \{\tilde{\mathbf{g}}\}$ ;
5:    $\tilde{\mathbf{g}} \leftarrow \text{Compose}(\tilde{\mathbf{g}}, F)$ ;
6: end while
7: return  $S$ ;

```

In Algorithm 1, the procedure “Compose” takes as input two sequences of polynomials $\mathbf{g} = (g_1, \dots, g_k)$ and $F = (f_1, \dots, f_n)$ in $\mathbb{Q}[x_1, \dots, x_n]$ and outputs a sequence of polynomials (h_1, \dots, h_k) in $\mathbb{Q}[x_1, \dots, x_n]$, such that $h_i = g_i(f_1, \dots, f_n)$ for all $1 \leq i \leq k$.

The procedure “InIdeal” takes as input two sequences of polynomials $\tilde{\mathbf{g}}$ and \mathbf{g} and decides if all the polynomials in $\tilde{\mathbf{g}}$ belong to the ideal generated by \mathbf{g} . It relies on the computation of a Gröbner basis for the ideal generated by \mathbf{g} , and the normal form of $\tilde{\mathbf{g}}$ w.r.t. to this basis [6, Chap 2, §6, Corollary 2].

We now prove the termination and correctness of Algorithm 1.

Theorem 2.4. *On input two sequences $\mathbf{g} = (g_1, \dots, g_k)$ and $F = (f_1, \dots, f_n)$ of polynomials in $\mathbb{Q}[x_1, \dots, x_n]$, Algorithm 1 terminates and outputs a sequence of polynomials whose vanishing set is the invariant set $S_{(F,V(\mathbf{g}))}$.*

PROOF. Consider the algebraic variety $V = V(\mathbf{g}) = V(g_1, \dots, g_k)$ and the polynomial map $F = (f_1, \dots, f_n) : \mathbb{C}^n \rightarrow \mathbb{C}^n$. Let $S_0 = \mathbf{g}$, and let S_m denote the set S after completing m iterations of the loop in Algorithm 1. Let I_m be the ideal generated by S_m . Since S_m is contained in S_{m+1} , the ideal I_m is contained in I_{m+1} . Therefore, we have the following ascending chain of ideals:

$$I_0 \subseteq I_1 \subseteq I_2 \subseteq \dots \subseteq I_m \subseteq I_{m+1} \subseteq \dots$$

Since $\mathbb{C}[x_1, \dots, x_n]$ is a Noetherian ring, the chain above becomes stationary which verifies that Algorithm 1 terminates.

To prove the correctness, let $X_m = \bigcap_{i=1}^m F^{-i}(X)$. We will prove that X_m is the algebraic variety associated to S_m for any m . By the construction of S_m , we have that:

$$S_m = \{g_1, \dots, g_k, g_1(F), \dots, g_k(F), \dots, g_1(F^m), \dots, g_k(F^m)\}$$

for any $m \in \mathbb{N}$. Using Lemma 2.1,

$$X_m = \bigcap_{i=0}^m F^{-i}(V(g_1, \dots, g_k)) = \bigcap_{i=0}^m V(g_1(F^i), \dots, g_k(F^i)) = V(S_m),$$

which implies that $X_m = V(I_m)$ for any m . The output of the algorithm is S_N such that $I_N = I_{N+1}$, since $\tilde{\mathbf{g}} \in I_N$. Therefore, $X_N = V(I_N) = V(I_{N+1}) = X_{N+1}$. By Theorem 2.3, $V(I_N) = X_N = S_{(F,X)}$, which implies the correctness of Algorithm 1. \square

3 GENERATING POLYNOMIAL LOOP INVARIANTS

We first fix our notation throughout this section. In the polynomial ring $\mathbb{C}[x_1, \dots, x_n]$, we fix the notation \mathbf{x}^α with $\alpha = (a_1, \dots, a_n) \in \mathbb{Z}_{\geq 0}^n$ denoting the monomial $x_1^{a_1} \dots x_n^{a_n}$. Throughout when we write $f = b_1 \mathbf{x}^{\alpha_1} + \dots + b_m \mathbf{x}^{\alpha_m}$, we refer to the expression of f in the basis of monomials, where f consists of exactly m terms (or monomials) $b_i \mathbf{x}^{\alpha_i}$ with coefficients $b_i \in \mathbb{C}$. In our polynomial expression, we always order the monomials such that for $i < j$:

$$\deg(\mathbf{x}^{\alpha_i}) < \deg(\mathbf{x}^{\alpha_j}) \text{ or } (\deg(\mathbf{x}^{\alpha_i}) = \deg(\mathbf{x}^{\alpha_j}) \text{ and } \mathbf{x}^{\alpha_i} >_{\text{lex}} \mathbf{x}^{\alpha_j})$$

where $\deg(\mathbf{x}^{\alpha_i})$ represents the degree of the monomial \mathbf{x}^{α_i} , and the lexicographic order is with respect to the order of the variables $x_1 > x_2 > \dots > x_n$. We also denote $|\alpha_i|$ for the size of the vector $\alpha_i = (\alpha_{i,1}, \dots, \alpha_{i,n})$ which is $\alpha_{i,1} + \dots + \alpha_{i,n}$.

3.1 The general case

Definition 3.1. *Let $\mathbf{a} \in \mathbb{C}^n$, and $\mathbf{g} = (g_1, \dots, g_k)$ and $F = (f_1, \dots, f_n)$ be two sequences of polynomials in $\mathbb{Q}[x_1, \dots, x_n]$. Consider the algebraic variety $X = V(\mathbf{g})$ and the polynomial map $F = (f_1, \dots, f_n)$. Then $\mathcal{L}(\mathbf{a}, \mathbf{g}, F)$ (or $\mathcal{L}(\mathbf{a}, X, F)$) denotes the following polynomial loop:*

```

 $(x_1, x_2, \dots, x_n) = (a_1, a_2, \dots, a_n)$ 
while  $g_1 = g_2 = \dots = g_k = 0$  do
   $\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \leftarrow F \begin{pmatrix} f_1 \\ f_2 \\ \vdots \\ f_n \end{pmatrix}$ 
end while

```

When no \mathbf{g} is identified, we will write $\mathcal{L}(\mathbf{a}, 0, F)$. Finally, we will simply write \mathcal{L} when it is clear from the context.

Proposition 3.2. *Let $\mathbf{a} \in \mathbb{C}^n$, X be an algebraic variety and $F : \mathbb{C}^n \rightarrow \mathbb{C}^n$ a polynomial map. Then, the polynomial loop $\mathcal{L}(\mathbf{a}, X, F)$ never terminates if and only if $\mathbf{a} \in S_{(F, X)}$.*

PROOF. The statement is direct from the definition, as $\mathcal{L}(\mathbf{a}, X, F)$ never terminates if, and only if, $F^{(m)}(\mathbf{a}) \in X$ for all $m \geq 0$ that is, if and only if $\mathbf{a} \in S_{(F, X)}$. \square

Example 1. Let us compute the termination condition for the following loop \mathcal{L} where $g = x_1^2 - x_1x_2 + 9x_1^3 - 24x_1^2x_2 + 16x_1x_2^2$.

```

(x1, x2) = (a1, a2)
while g = 0 do
  (x1, x2) ← (10x1 - 8x2, 6x1 - 4x2)
end while

```

Algorithm 1 computes the invariant set $S_{(F, X)}$ of the variety $X = V(g)$ for the map $F = (f_1, f_2) = (10x_1 - 8x_2, 6x_1 - 4x_2)$ as follows:

$$S_{(F, X)} = V(x_1 - x_2 - 9x_1^2 + 24x_1x_2 - x_2^2).$$

Therefore, by Proposition 3.2 we have that \mathcal{L} never terminates if and only if $(a_1, a_2) \in V(x_1 - x_2 - 9x_1^2 + 24x_1x_2 - x_2^2)$.

Definition 3.3. *Polynomial invariants of a loop \mathcal{L} are polynomials that vanish before and after every iteration of \mathcal{L} . The set $I_{\mathcal{L}}$ of all polynomial invariants for \mathcal{L} is an ideal, called the invariant ideal of \mathcal{L} . Let $d \geq 1$, the subset $I_{d, \mathcal{L}}$ of all polynomial invariant for \mathcal{L} , of total degree $\leq d$, is called the d^{th} truncated invariant ideal of \mathcal{L} .*

Though a truncated invariant ideal is not an ideal, it has the structure of a finite dimensional vector space. Hence, it can be uniquely parametrized by a system of linear equations, whose coefficients depend on the initial values. In the following, we demonstrate how to reduce the computation of such a parametrization for a given loop to computing an invariant set of an extended polynomial map.

We start by a criterion to determine whether a given polynomial is invariant with respect to a given loop or not.

Proposition 3.4. *Let $\mathbf{a} \in \mathbb{C}^n$ and $F = (f_1, \dots, f_n) \subset \mathbb{C}[x_1, \dots, x_n]$. For $m = \binom{n+d}{d}$, let $\mathbf{b} \in \mathbb{C}^m$ and*

$$g(\mathbf{x}, \mathbf{y}) := \sum_{|\alpha_i| \leq d} y_i \mathbf{x}^{\alpha_i}$$

be a degree d polynomial in $\mathbb{C}[x_1, \dots, x_n, y_1, \dots, y_m]$. Then $g(\mathbf{x}, \mathbf{b})$ is a polynomial invariant for $\mathcal{L}(\mathbf{a}, 0, F)$ if, and only if, $(\mathbf{a}, \mathbf{b}) \in S_{(F_m, X)}$ where $F_m = (f_1, \dots, f_n, y_1, \dots, y_m)$ and $X = V(g) \subset \mathbb{C}^{n+m}$.

PROOF. Consider the following “ m^{th} extended” loop $\mathcal{L}((\mathbf{a}, \mathbf{b}), g, F_m)$:

```

(x, y) = (a, b)
while g(x, y) = 0 do
  (x1, ..., xn, y1, ..., ym) ← (f1(x1, ..., xn), ..., fn(x1, ..., xn), y1, ..., ym)
end while

```

Let $\mathbf{a}^0 = \mathbf{a}$ and for $k \geq 1$ let $\mathbf{a}^k = F(\mathbf{a}^{k-1})$. Then, $(\mathbf{a}^k)_{k \in \mathbb{N}}$ are the successive values of \mathbf{x} in $\mathcal{L}(\mathbf{a}, 0, F)$. Let $\mathbf{b} \in \mathbb{C}^m$ and assume that $g(\mathbf{x}, \mathbf{b})$ is a polynomial invariant for $\mathcal{L}(\mathbf{a}, 0, F)$. Let $k \geq 0$, then after the k^{th} iteration of the extended loop $\mathcal{L}((\mathbf{a}, \mathbf{b}), g, F_m)$, the value of \mathbf{x} is \mathbf{a}^k and the value of \mathbf{y} is still \mathbf{b} . Since, by assumption $g(\mathbf{a}^k, \mathbf{b}) = 0$, this loop does not stop after the k^{th} iteration and, by induction never terminates. The converse is immediate.

Therefore, $g(\mathbf{x}, \mathbf{b})$ is a polynomial invariant for \mathcal{L} if and only if the extended loop \mathcal{L}_m never terminates, which is equivalent, by Proposition 3.2, to $(\mathbf{a}, \mathbf{b}) \in S_{(F_m, V(g))}$. \square

The following main result follows from the above criterion.

Theorem 3.5. *Let $F = (f_1, \dots, f_n)$ be a sequences of polynomials in $\mathbb{Q}[x_1, \dots, x_n]$ and let $d \geq 1$ and $m = \binom{n+d}{d}$. Then, there exists an algorithm *TruncatedInvariant* which, on input (F, d) computes a polynomial matrix A , with m columns, and coefficients in $\mathbb{C}[x_1, \dots, x_n]$, such that the d^{th} truncated invariant ideal of $\mathcal{L}(\mathbf{a}, 0, F)$ for any $\mathbf{a} \in \mathbb{Q}^n$ is equal to:*

$$I_{d, \mathcal{L}} = \left\{ \sum_{|\alpha_i| \leq d} b_i \mathbf{x}^{\alpha_i} \mid (b_1, \dots, b_m) \in \ker A(\mathbf{a}) \right\}$$

where $\ker A(\mathbf{a})$ is right-kernel of A , whose entries are evaluated at \mathbf{a} .

PROOF. Let y_1, \dots, y_m be new indeterminates, and define g, F_m and X as in Proposition 3.4. Then, by Proposition 2.4, on input (g, F_m) , Algorithm 1 computes polynomials h_1, \dots, h_N in $\mathbb{Q}[x_1, \dots, x_n, y_1, \dots, y_m]$, whose common vanishing set is $S_{(F_m, X)}$. Moreover, by construction of Algorithm 1 and definition of F_m , we have:

$$h_j = g \circ F_m^j(x_1, \dots, x_n, y_1, \dots, y_m) = g(F^j(x_1, \dots, x_n), y_1, \dots, y_m)$$

for $0 \leq j \leq N$. Thus, the h_j 's are linear in the variables y_i 's. Then, there exists a matrix A , with coefficients in $\mathbb{C}[x_1, \dots, x_n]$ and with m columns, such that

$$\begin{bmatrix} h_1 \\ \vdots \\ h_m \end{bmatrix} = A \cdot \begin{bmatrix} y_1 \\ \vdots \\ y_m \end{bmatrix} \quad (1)$$

Let $\mathbf{b} \in \mathbb{Q}^n$, by Proposition 3.4, $g(\mathbf{x}, \mathbf{b})$ is a polynomial invariant of $\mathcal{L}(\mathbf{a}, 0, F)$ if, and only if, $(\mathbf{a}, \mathbf{b}) \in S_{F_m, X}$, that is, by (1), if and only if $A(a_1, \dots, a_n) \cdot \mathbf{b} = 0$. Since any polynomial in $\mathbb{Q}[x_1, \dots, x_n]$ can be written as $g(\mathbf{x}, \mathbf{b})$, for some $\mathbf{b} \in \mathbb{Q}^m$, we are done. \square

Example 2. We consider the following polynomial loop \mathcal{L} :

```

(x1, x2) = (a1, a2)
while true do
  (x1, x2) ← (10x1 - 8x2, 6x1 - 4x2)
end while

```

We proceed to compute the second truncated polynomial ideal for \mathcal{L} using the algorithm outlined in the proof of Theorem 3.5. Some of the polynomial invariants for this loop have been computed in [14] for specific initial values, and are used to verify the non-termination of the linear loop with the assignment “ $2x_2 - x_1 \geq -2$ ”. In our analysis, we extend this validation by computing *all* polynomial invariants up to degree 2 for *arbitrary* initial value.

We first run Algorithm 1 on input $F_6 = (10x_1 - 8x_2, 6x_1 - 4x_2, y_1, \dots, y_6)$, and $g = y_1 + y_2x_1 + y_3x_2 + y_4x_1^2 + y_5x_1x_2 + y_6x_2^2$ where the y_i 's are new variables. The output is polynomials h_1, \dots, h_5 in $\mathbb{C}[x_1, x_2, y_1, \dots, y_6]$ whose common zero set is $S_{(F_6, X)} \subset \mathbb{C}^8$.

As the h_i 's are linear in the y_j 's, we can write them as the product of the matrix

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3x_1 - 4x_2 & 3x_1 - 4x_2 & 0 & 0 & 0 \\ 0 & 64x_2 & 112x_2 - 48x_1 & 48x_2^2 & 84x_2^2 - 36x_1x_2 & 27x_1^2 - 126x_1x_2 + 147x_2^2 \\ 0 & 32x_2 & 56x_2 - 24x_1 & 24x_1x_2 & -9x_1^2 + 21x_1x_2 + 12x_2^2 & -18x_1x_2 + 42x_2^2 \\ 0 & 4x_2 & 7x_2 - 3x_1 & 3x_1^2 & 3x_1x_2 & 3x_2^2 \end{bmatrix}$$

by the vector whose entries are the y_1, \dots, y_6 . This matrix is the output of the procedure `TruncatedInvariant` given in Theorem 3.5. Here, we actually show a reduced version of this matrix for clarity reasons: we used the “trim” command from Macaulay [13], to find smaller generators for the ideal generated by the g_i 's.

Actually, from this output we can go further by computing an explicit basis for the corresponding vector space of $I_{2, \mathcal{L}}$. This is done by computing a basis for the kernel of the above matrix, depending of the possible values for (a_1, a_2) . Performing Gauss elimination on this matrix, we are led to consider the following four cases:

Initial values	Basis of $I_{2, \mathcal{L}}$
$a_1 = a_2 = 0$	$\{x_1, x_2, x_1x_2, x_1^2, x_2^2\}$
$a_1 = a_2 \neq 0$	$\{x_1 - x_2, x_1^2 - x_1x_2, -x_1x_2 + x_2^2\}$
$a_1 = \frac{4}{3}a_2 \neq 0$	$\{3x_1 - 4x_2, -3x_1^2 + 16x_1x_2 - 16x_2^2, -3x_1x_2 + 4x_2^2\}$
$a_1 \neq \frac{4}{3}a_2, a_1 \neq a_2$	$\{(3a_1 - 4a_2)^2x_1 - (3a_1 - 4a_2)^2x_2 - 9(a_1 - a_2)x_1^2 + 24(a_1 - a_2)x_1x_2 - 16(a_1 - a_2)x_2^2\}$

It is remarkable that in the first three cases, the truncated invariant ideal does not depend on the initial value. This is because they correspond to degenerate cases where the initial values are not generic; that is, they lie in a proper algebraic variety of \mathbb{C}^2 . However, the last case is generic, and the output depends on the initial values. This is the output one would obtain by running Gauss elimination on the above polynomial matrix in the field of rational fractions in the x_i 's. However, such a computation is not tractable in general, as the size of the expressions increases quickly.

3.2 Loops with given initial value

While the algorithm outlined Theorem 3.5 tackles the most general case, in practice it quickly becomes unpractical, even for small inputs. In this section, we focus on the particular case where the initial values of the loops are fixed, and design an adapted efficient algorithm for this case. We will see in Section 4.2 that the solution to this particular problem can be used to partially solve the general problem.

The subsequent proposition outlines a sufficient condition for a polynomial to qualify as an invariant polynomial, utilizing the initial values of the loop.

Proposition 3.6. *Consider a loop $\mathcal{L}(\mathbf{a}_0, 0, F)$. Let $\mathbf{a}_n = F^{(n)}(\mathbf{a}_0)$. If $\sum_{i=1}^m y_i \mathbf{x}^{\alpha_i}$ is a polynomial invariant, then y_i 's satisfy the equations:*

$$\sum_{i=1}^m y_i \mathbf{a}_0^{\alpha_i} = \dots = \sum_{i=1}^m y_i \mathbf{a}_k^{\alpha_i} = 0.$$

Proposition 3.6 is a direct consequence of the following lemma.

Lemma 3.7. *Let $\mathbf{a}_0 \in \mathbb{C}^n$ and $F = (f_1, \dots, f_n) \in \mathbb{C}[x_1, \dots, x_n]$. Let*

$$X = V\left(\sum_{i=1}^m y_i \mathbf{x}^{\alpha_i}\right) \subset \mathbb{C}^{n+m}.$$

For all $k \in \mathbb{N}$, let $\mathbf{a}_k = F^{(k)}(\mathbf{a}_0)$ and let

$$X_k = \bigcap_{j=0}^k F_m^{-j}(X) \quad \text{and} \quad S_k = X_k \cap V(\mathbf{x} - \mathbf{a}_0)$$

. Then, the following holds:

$$(a) \quad S_k = V\left(\sum_{i=1}^m y_i \mathbf{a}_0^{\alpha_i}, \dots, \sum_{i=1}^m y_i \mathbf{a}_k^{\alpha_i}, \mathbf{x} - \mathbf{a}_0\right)$$

$$(b) \quad S_{(F_m, X)} \cap V(\mathbf{x} - \mathbf{a}_0) \subset S_k \text{ for any } k \in \mathbb{N}.$$

PROOF. (a) Since $F_m = (f_1, \dots, f_n, y_1, \dots, y_m)$ then for $j \geq 0$, we can note $F_m^{(j)} = (f_{j,1}, \dots, f_{j,n}, y_1, \dots, y_m)$. Then, according to Lemma 2.1, we can rewrite X_k as

$$\bigcap_{j=0}^k F_m^{-j}\left(V\left(\sum_{i=1}^m y_i \mathbf{x}^{\alpha_i}\right)\right) = V\left(\sum_{i=1}^m y_i \mathbf{x}^{\alpha_i}, \dots, \sum_{i=1}^m y_i (F^k(\mathbf{x}))^{\alpha_i}\right),$$

so that

$$\begin{aligned} S_k &= V\left(\sum_{i=1}^m y_i \mathbf{x}^{\alpha_i}, \dots, \sum_{i=1}^m y_i (F^k(\mathbf{x}))^{\alpha_i}, \mathbf{x} - \mathbf{a}_0\right) \\ &= V\left(\sum_{i=1}^m y_i \mathbf{a}_0^{\alpha_i}, \dots, \sum_{i=1}^m y_i \mathbf{a}_k^{\alpha_i}, \mathbf{x} - \mathbf{a}_0\right). \end{aligned}$$

(b) By Proposition 2.3, we have the following descending chain:

$$X_0 \supset X_1 \supset \dots \supset X_N = S_{(F_m, X)} = X_{N+1} = \dots \text{ for some } N \in \mathbb{N}.$$

Thus, by intersecting the above chain with an algebraic variety $V = V(x_1 - a_{0,1}, \dots, x_n - a_{0,n})$ we get the descending chain:

$$S_0 \supset S_1 \supset \dots \supset S_N = S_{(F_m, X)} \cap V = S_{N+1} = \dots \text{ for some } N \in \mathbb{N}.$$

Thus, $S_{(F_m, X)} \cap V$ is a subset of S_k for any $k \in \mathbb{N}$. \square

Therefore, when a loop's initial value is set, Proposition 3.6 provides arbitrary many linear equations M , which may be linearly dependent, for the coefficients of a degree d polynomial invariant. This accelerate polynomial invariant computations by providing a solution set containing the desired truncated invariant ideal so that a possible choice for M is $\binom{n+d}{d}$, the dimension of this truncated ideal. In particular, a vector basis \mathcal{B} of this solution set serves as candidates for the basis of the truncated invariant ideal.

Following the above idea, we now describe an algorithm for computing a vector basis of the d^{th} truncated invariant ideal of a loop, with a fixed initial value. We first explain two subroutines used in this algorithm.

The procedure `VectorBasis` takes linear forms as input and computes a vector basis of the common vanishing set these forms.

The procedure `CheckPI` takes as input $\mathbf{a} \in \mathbb{Q}^n$ and polynomials $F = (f_1, \dots, f_n)$ and g in $\mathbb{Q}[x_1, \dots, x_n]$. It outputs True if, and only, if g is a polynomial invariant of $\mathcal{L}(\mathbf{a}, 0, F)$. Such a procedure can be obtained by a direct combination of an application of Algorithm 1 to (g, F) and the effective criterion given by Proposition 3.2.

Algorithm 2 Computing truncated invariant ideals

Input: A sequence of rational numbers $\mathbf{a} = (a_1, \dots, a_n)$, a natural number d and a sequence of polynomials $F = (f_1, \dots, f_n) \in \mathbb{Q}[x_1, \dots, x_n]$.

Output: Polynomials forming a vector space basis for the d^{th} truncated ideal of the loop $\mathcal{L}(\mathbf{a}, 0, F)$.

```

1:  $g \leftarrow \sum_{|\alpha_i| \leq d} y_i x^{\alpha_i}$ ;
2:  $M \leftarrow \binom{n+d}{d}$ ;
3:  $(b_1, \dots, b_m) \leftarrow \text{VectorBasis}(g(\mathbf{a}, \mathbf{y}), g(F(\mathbf{a}), \mathbf{y}), \dots, g(F^M(\mathbf{a}), \mathbf{y}));$ 
4:  $\mathcal{B} \leftarrow (\sum_{|\alpha_i| \leq d} b_{1,i} x^{\alpha_i}, \dots, \sum_{|\alpha_i| \leq d} b_{m,i} x^{\alpha_i})$ ;
5:  $C = (h_1, \dots, h_l) \leftarrow \{h \in \mathcal{B} \mid \text{CheckPI}(\mathbf{a}, F, h) == \text{False}\}$ ;
6: if  $C == \emptyset$ , then
7:   return  $\mathcal{B}$ ;
8: else
9:    $(\tilde{h}_1, \dots, \tilde{h}_k) \leftarrow \text{InvariantSet}(\sum_{j=1}^l z_j h_j, (f_1, \dots, f_n, z_1, \dots, z_l))$ ;
10:   $(b'_1, \dots, b'_s) \leftarrow \text{VectorBasis}(\tilde{h}_1(\mathbf{a}, \mathbf{z}), \dots, \tilde{h}_k(\mathbf{a}, \mathbf{z}))$ ;
11:   $\mathcal{B}_1 \leftarrow (\sum_{j=1}^l b'_{1,i} h_j, \dots, \sum_{j=1}^l b'_{s,i} h_j)$ ;
12:   $\mathcal{B}_2 \leftarrow \mathcal{B}.\text{remove}(C)$ ;
13:   $\mathcal{B} \leftarrow \mathcal{B}_1.\text{extend}(\mathcal{B}_2)$ ;
14: return  $\mathcal{B}$ ;
15: end if
```

We now prove the correctness of Algorithm 2.

Theorem 3.8. *On input a sequence of $\mathbf{a} = (a_1, \dots, a_n)$ in \mathbb{Q}^n , a sequence of polynomials $F = (f_1, \dots, f_n) \in \mathbb{Q}[x_1, \dots, x_n]$ and $d \in \mathbb{N}$, Algorithm 2 outputs a sequence of polynomials which is a basis for the d^{th} truncated ideal for the loop $\mathcal{L}(\mathbf{a}, 0, F)$.*

PROOF. Assume that $g = \sum_{|\alpha_i| \leq d} y_i x^{\alpha_i}$ is a polynomial invariant for \mathcal{L} . Let $\{b_1, \dots, b_m\}$ be a basis for the solution set of $g(\mathbf{a}, \mathbf{y}) = g(F(\mathbf{a}), \mathbf{y}) = \dots = g(F^M(\mathbf{a}), \mathbf{y}) = 0$ where $M = \binom{n+d}{d}$. Then,

$$\mathcal{B} = \{ \sum_{|\alpha_i| \leq d} b_{1,i} x^{\alpha_i}, \dots, \sum_{|\alpha_i| \leq d} b_{m,i} x^{\alpha_i} \}$$

consists of linearly independent polynomials in $\mathbb{C}[x_1, \dots, x_n]_{\leq d}$. By Proposition 3.6, the variables \mathbf{y} satisfy linear equations $g(\mathbf{a}, \mathbf{y}) = g(F(\mathbf{a}), \mathbf{y}) = \dots = g(F^M(\mathbf{a}), \mathbf{y}) = 0$. Therefore, $I_{d,\mathcal{L}}$ is contained in the vector space generated by \mathcal{B} . Let $C = \{h_1, \dots, h_l\}$ be the set of all polynomials in \mathcal{B} that are not polynomial invariants. Assume that C is not empty. Otherwise, every polynomial in \mathcal{B} is a polynomial invariant for \mathcal{L} which implies that \mathcal{B} is a basis for $I_{d,\mathcal{L}}$.

By Proposition 3.4, $\sum_{j=1}^l z_j h_j$ is a polynomial invariant if and only if $\tilde{h}_1(\mathbf{a}, \mathbf{z}) = \dots = \tilde{h}_k(\mathbf{a}, \mathbf{z}) = 0$, where

$$(\tilde{h}_1, \dots, \tilde{h}_k) = \text{InvariantSet}(\sum_{j=1}^l z_j h_j, (f_1, \dots, f_n, z_1, \dots, z_l)).$$

Since $\tilde{h}_1(\mathbf{a}, \mathbf{z}) = \dots = \tilde{h}_k(\mathbf{a}, \mathbf{z}) = 0$ represents a system of linear equations, we can find a basis \mathcal{B}_1 for a subspace V_1 that is the intersection of $I_{d,\mathcal{L}}$ and the vector space generated by C , using exactly the same method employed for computing \mathcal{B} . Denote the set $\mathcal{B} \setminus C = \{g_1, \dots, g_{m-l}\}$ by \mathcal{B}_2 .

Now, we will show that the vector space V generated by $\mathcal{B}_1 \cup \mathcal{B}_2$ is equal to $I_{d,\mathcal{L}}$. Since $\mathcal{B}_1 \cup \mathcal{B}_2 \subseteq I_{d,\mathcal{L}}$, it follows that V is a subset of $I_{d,\mathcal{L}}$. To prove the other inclusion, let $g \in I_{d,\mathcal{L}}$. Since $I_{d,\mathcal{L}}$ is contained in the vector space generated by \mathcal{B} , there exist $c_1, \dots, c_m \in \mathbb{C}$ such that $g = \sum_{i=1}^l c_i \tilde{h}_i + \sum_{j=1}^{m-l} c_{l+j} g_j$. Then, $g - \sum_{j=1}^{m-l} c_{l+j} g_j$ is a polynomial invariant given that g, g_{l+1}, \dots, g_m are all polynomial invariants. Thus, $\sum_{i=1}^l c_i g_i$ is a polynomial invariant contained in the vector

space generated by C , implying $\sum_{i=1}^l c_i g_i \in V'$. Hence, $g \in V$ and so, $I_{d,\mathcal{L}} \subseteq V$. Since the intersection of the vector space generated by \mathcal{B}_1 and the vector space generated by \mathcal{B}_2 is $\{0\}$, we conclude that $\mathcal{B}_1 \cup \mathcal{B}_2$ is a basis for $I_{d,\mathcal{L}}$, which completes the proof. \square

We now compute all polynomial invariants for the loop in Example 3 up to degree 4 using Algorithm 2.

Example 3 (Squares). The d^{th} truncated invariant ideals of the ‘‘Squares’’ loop in Appendix A. cannot be computed by Algorithm 1 within an hour, for $d = 2, 3, 4$. However, Algorithm 2 easily computes them. The input for Algorithm 2 to compute $I_{2,\mathcal{L}}$ is $(-1, -1, 1)$, 2, and F . Assume that $g = y_1 + y_2 x_1 + \dots + y_{10} x_3^2$ is a polynomial invariant of degree 2. By Proposition 3.6, this leads to 10 linear equations, whose solutions provide the following candidates for polynomial invariants:

$$\begin{aligned} \mathcal{B} = & \{1 + x_1 + x_2 + x_3, 1 + x_1 + x_2 + x_3^2, 2 + 3x_1 + 3x_2 \\ & + x_1^2 + 2x_1 x_2 + x_2^2, -2 - x_1 - 3x_2 + x_1^2 + 2x_1 x_3 - x_2^2, \\ & 2 - 3x_1 - x_2 - x_1^2 + x_2^2 + 2x_2 x_3\}. \end{aligned}$$

The procedure CheckPI verifies that all the polynomials in \mathcal{B} are invariant polynomials, and \mathcal{B} forms a basis for $I_{2,\mathcal{L}}$. Moreover, the outputs of Algorithm 2 show that $I_{3,\mathcal{L}}$ represents a 13-dimensional vector space, and $I_{4,\mathcal{L}}$ a 26-dimensional vector space.

This example has been previously explored in [1], where only a closed formula is derived as $x(n) + y(n) = 2^n(x(0) + y(0) + 2) - (-1)^n/2 - 3/2$. We calculate truncated invariant ideals $I_{d,\mathcal{L}}$ for $d = 1, 2, 3, 4$, considering a given initial value. This covers all polynomial invariants up to degree 4 for Example 3 using Algorithm 2.

4 APPLICATIONS AND FURTHER RESULTS

In this section, we show different applications of Algorithm 2 and their consequences to various examples from the literature.

4.1 On the Fibonacci sequences

The following examples, discussed in [1], are significant in the theory of trace maps, see e.g. [3, 24]. In each example, Algorithm 2 computes truncated polynomial ideals up to degree 4, establishing that in each case, there are no polynomial invariants up to degree 2.

In Example 4, we prove that the computed polynomial invariant of degree 3 by Algorithm 2, generates the entire invariant ideal.

Example 4 (Fibonacci sequence). The Fibonacci numbers follow the recurrence relation: $F_0 = 0$, $F_1 = 1$, and $F_n = F_{n-1} + F_{n-2}$ for all $n \geq 2$. We can express the Fibonacci sequence as a loop \mathcal{F} .

```

 $(x_1, x_2) = (0, 1)$ 
while true do
   $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \leftarrow \begin{pmatrix} x_2 \\ x_1 + x_2 \end{pmatrix}$ 
end while

```

Algorithm 2 computes that the truncated invariant ideals $I_{2,\mathcal{F}}$ and $I_{3,\mathcal{F}}$ are zero, and $g = -1 + x_1^4 + 2x_1^3x_2 - x_1^2x_2^2 - 2x_1x_2^3 + x_2^4$ forms a basis for $I_{4,\mathcal{F}}$. Therefore, the Fibonacci numbers satisfy the equation:

$$F_{n-1}^4 + 2F_{n-1}^3F_n - F_{n-1}^2F_n^2 - 2F_{n-1}F_n^3 + F_n^4 - 1 = 0 \text{ for all } n \in \mathbb{N}.$$

Moreover, $I_{\mathcal{F}}$ is generated by g . To prove that, observe that

$$g = (-1 - x_1^2 - x_1x_2 + x_2^2)(1 - x_1^2 - x_1x_2 + x_2^2)$$

and (F_{n-1}, F_n) lies on $V(1 - x_1^2 - x_1x_2 + x_2^2)$ for infinitely many n . Thus, for any $f \in I_{\mathcal{F}}$, the system of equations $1 - x_1^2 - x_1x_2 + x_2^2 = f(x_1, x_2) = 0$ has infinitely many solutions. By [28, Page 4], $f(x_1, x_2)$ is divisible by $1 - x_1^2 - x_1x_2 + x_2^2$. The same applies to $-1 - x_1^2 - x_1x_2 + x_2^2$. Consequently, $f(x_1, x_2)$ is divisible by g , establishing that $I_{\mathcal{F}}$ is generated by g .

For the following examples, Algorithm 2 computes a unique polynomial invariant in degree 3. Proposition 3.4 and Algorithm 2 demonstrate that the identified polynomial is the sole invariant of degree 3, and all degree 4 polynomials can be derived from it. The proof of uniqueness is a novel contribution. Additionally, the polynomials given in [2] for Fib2 and Fib3 were found to be incorrect.

Example 5. For the Fib1, Fib2 and Fib3 loops from Appendix A Algorithm 2 computes a basis for the truncated invariant ideals as:

$$I_{1,\mathcal{L}} = I_{2,\mathcal{L}} = \{0\}, \quad I_{3,\mathcal{L}} = \{g\}, \quad \text{and} \quad I_{4,\mathcal{L}} = \{g, x_1g, x_2g, x_3g\},$$

where g for Fib1, Fib2 and Fib3 is, respectively,

- $-2 + x_1^2 + x_2^2 + x_3^2 - 2x_1x_2x_3$
- $76 - x_2 - 2x_1x_3 + 4x_1^2x_2$
- $7 + x_1 + x_2 + x_3 - x_1^2 + x_1x_2 + x_1x_3 - x_2^2 + x_2x_3 - x_3^2 + x_1x_2x_3$.

Note that a basis for $I_{4,\mathcal{L}}$ is generated by a basis for $I_{3,\mathcal{L}}$.

4.2 Invariant lifting for generic initial values

In this section, we introduce a method that computes a polynomial invariant for any initial value from a polynomial invariant for a specific initial value. Assume that we have a polynomial invariant f for a loop \mathcal{L} with a given initial value. Then, our method checks whether $f - f(a)$ is a polynomial invariant for \mathcal{L} for any initial value a . Moreover, $f - h(a)$ is a polynomial invariant for \mathcal{L} for any initial value a if and only if $h = f$.

Proposition 4.1. A polynomial invariant $f(\mathbf{x}) = 0$ for a loop \mathcal{L} with given initial value and a polynomial map F can be extended to a polynomial invariant $f(\mathbf{x}) - f(\mathbf{a}) = 0$ for \mathcal{L} with any initial value a if and only if $S_{(F_1,X)} = X$, where $X = V(f(\mathbf{x}) - t)$.

PROOF. Assume that $f(\mathbf{x}) - f(\mathbf{a}) = 0$ is a polynomial invariant for \mathcal{L} for any initial value a . Therefore, \mathcal{L}_1 with a guard $f(\mathbf{x}) - t = 0$ never terminates if $t = f(\mathbf{a})$. Hence, $X = V(t - f(\mathbf{x})) \subset S(F_1, X)$

and $S_{(F_1,X)} \subset X$ by Proposition 2.3, implying that $X = V(S_{F_1,X})$. To prove the other direction, assume $S_{(F_1,X)} = X$. By the definition of the invariant set, \mathcal{L}_1 with a guard $f(\mathbf{x}) - t = 0$ never terminates if and only if initial value of (\mathbf{x}, t) is contained in $S_{(F_1,X)}$. Therefore, \mathcal{L}_1 never terminates if and only if $t = f(\mathbf{a})$ for any initial value a . By Proposition 3.4, $f(\mathbf{x}) - f(\mathbf{a}) = 0$ is a polynomial invariant for \mathcal{L} with any initial value a . \square

Example 6. Consider the loops from Example 5. We compute polynomial invariants of the form $f(x_1, x_2, x_3) - f(a_1, a_2, a_3) = 0$. Algorithm 1, with the input being a polynomial map F_1 and the algebraic variety $X = V(x_1^2 + x_2^2 + x_3^2 - 2x_1x_2x_3 - 2 - t) \subset \mathbb{C}^4$, computes $S_{(F_1,X)}$, which is equal to $V(x_1^2 + x_2^2 + x_3^2 - 2x_1x_2x_3 - 2 - t)$. Then, $x_1^2 + x_2^2 + x_3^2 - 2x_1x_2x_3 - (a_1^2 + a_2^2 + a_3^2 - 2a_1a_2a_3) = 0$ is a general polynomial invariant for Fib1, Example 5, as stated in Proposition 4.1. Similarly, we verify that polynomial invariants of Fib2 and Fib3 can be extended to a general polynomial invariant of the form $f(x_1, x_2, x_3) - f(a_1, a_2, a_3) = 0$.

The following example from [29], gained significance in the context of the Jacobian conjecture, as discussed in [1], since it illustrates a lack of linear invariants, serving as a counterexample to the linear dependence conjecture within Yagzhev maps [8].

In Example 7, Algorithm 2 computes the first truncated invariant ideal, but only some polynomials in the second truncated ideal.

Example 7 (Yagzhev9). Consider the loop “Yagzhev9” from Appendix A. Algorithm 2 computes $\{-7 - x_7 + x_8, x_2 - x_4 + x_6, x_1 - x_3 + x_5\}$ as a basis for $I_{1,\mathcal{L}}$. Additionally, it generates the degree 2 polynomials:

$$11 - x_1x_4 + x_2x_3, \quad 11 - x_3x_6 + x_4x_5 \text{ and } 11 - x_1x_6 + x_2x_5.$$

By Algorithm 1 and Proposition 4.1, we verify that the polynomials:

$$-x_1x_4 + x_2x_3 - (-a_1a_4 + a_2a_3) = 0;$$

$$-x_3x_6 + x_4x_5 - (-a_3a_6 + a_4a_5) = 0;$$

$$-x_1x_6 + x_2x_5 - (-a_1a_6 + a_2a_5) = 0.$$

are general polynomial invariant for any initial value (a_1, \dots, a_9) . However, the polynomial invariants in $I_{1,\mathcal{L}}$ cannot be extended to general polynomial invariants.

4.3 A sufficient criterion for termination of semi-algebraic loops

Definition 4.2. Consider the basic semi-algebraic set S of \mathbb{R}^n defined by $g_1 = \dots = g_k = 0$ and $h_1 > 0, \dots, h_s > 0$ and a polynomial map $F = (f_1, \dots, f_n)$, where the f_i 's, the g_j 's and the h_j 's are polynomials in $\mathbb{R}[x_1, \dots, x_n]$. Then a loop of the form:

```

 $(x_1, x_2, \dots, x_n) = (a_1, a_2, \dots, a_n)$ 
while  $g_1 = \dots = g_k = 0$  and  $h_1 > 0, \dots, h_s > 0$  do
   $\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \leftarrow \begin{pmatrix} f_1 \\ f_2 \\ \vdots \\ f_n \end{pmatrix}$ 
end while

```

is called a semi-algebraic loop on S with respect to F .

We denote by $S(\mathbf{g}, \mathbf{h})$ the solution set in \mathbb{R}^n of the polynomial system defined by \mathbf{g} and \mathbf{h} .

The following proposition is a direct consequence of the definitions.

Proposition 4.3. *Let $\mathbf{a} \in \mathbb{Q}^n$, and let \mathbf{g} and \mathbf{h} be as in Definition 4.2. Let r_1, \dots, r_p be polynomial invariants of $\mathcal{L}(\mathbf{a}, 0, F)$, then the semi-algebraic loop $\mathcal{L}(\mathbf{a}, (\mathbf{g}, \mathbf{h}), F)$ never terminates if*

$$V(r_1, \dots, r_p) \cap \mathbb{R}^n \subset S(\mathbf{g}, \mathbf{h}).$$

The above inclusion **correspond** to the quantified formula:

$$\forall \mathbf{x} \in \mathbb{R}^n, r_1(\mathbf{x}) = \dots = r_p(\mathbf{x}) = 0 \Rightarrow \begin{cases} g_1(\mathbf{x}) = \dots = g_k(\mathbf{x}) = 0 \\ h_1(\mathbf{x}) > 0, \dots, h_s(\mathbf{x}) > 0 \end{cases}.$$

The validity of such **formula** can be **deciding** using a quantifier elimination algorithm [4, Chapter 14]. Since there is no quantifier alternate, it actually correspond to the emptiness decision of solution set for a polynomial system of equation and inequalities. This is efficiently tackled by specific algorithms, whose most general version can be found in [4, Theorem 13.24]. Besides, given the particular structure of this formula, an efficient approach would be to follow the one of [12], which is base on a combination of the Real Nullstellensatz [5] and Putinar’s Positivstellensatz [23].

We do not go further on these aspects as it falls behind the goal of this paper and these directions will be explored in future works. Instead, we show in the following example, why the above sufficient criterion is not a necessary one.

Example 8. Consider the elementary semi-algebraic loop:

$$\begin{aligned} &(x_1, x_2) = (a_1, a_2) \\ &\textbf{while } x_1 > 0 \textbf{ do} \\ &\quad \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \leftarrow \begin{pmatrix} 2x_1 \\ 2x_2 \end{pmatrix} \\ &\textbf{end while} \end{aligned}$$

A direct study of the linear recursive sequence defined by the successive values $\mathbf{a}^0, \mathbf{a}^1, \dots$ of (x_1, x_2) shows that this loop never terminates if, and only if $a_1 > 0$. Besides, $a_2x_1 - a_1x_2 = 0$ is a polynomial invariant of this loop, and since every \mathbf{a}^j , for $j \geq 0$ must be on this line, it generates the whole invariant ideal. However, $V(a_2x_1 - a_1x_2) \cap \mathbb{R}^2$ is not contained in $S(0, x_1)$.

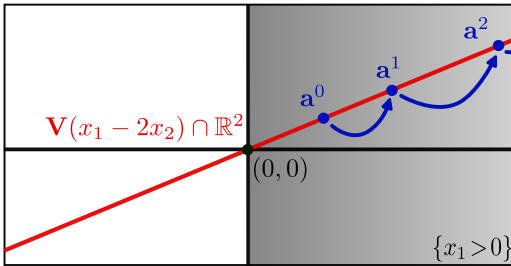


Figure 1: An illustration of a particular case of Example 8 where $(a_1, a_2) = (2, 1)$. In blue are depicted the successive values $\mathbf{a}^0, \mathbf{a}^1, \dots$ of the variables (x_1, x_2) , in red is the real zero-set of the invariant ideal, and in gray the set $S(0, x_1)$ defined by the condition $x_1 > 0$.

5 IMPLEMENTATION AND EXPERIMENTAL RESULTS

In this section, we present an implementation of the algorithms presented in this paper, and we discuss its performances on some

examples from the literature. We also compare it to the algorithm from [1]. Example 2 is sourced from [14], “Floor function” is cited from [25], while the remaining examples are from [2].

5.1 Implementation details

We implemented Algorithm 1 and Algorithm 2 in Macaulay [13] and performed the experiments on a laptop equipped with a 4.8 GHz Intel i7 processor, 16 GB of RAM and 25 MB L3 cache. In particular, the subroutines mainly rely on classic linear algebra routines, and implemented G  bner bases algorithms.

The implementation we wrote benefits from some speedup which comes from heuristics deduced from several experiments.

First, we observed that most of the time, for simple loops, all the candidate polynomials in \mathcal{B} , computed at step 4 of Algorithm 2, are actually polynomial invariants. Another observation, is that the smaller is the dimension of the variety X , the faster Algorithm 1 computes polynomials defining $S_{F,X}$, for some polynomial map F . Hence, instead of checking individually, we decide first if all elements of \mathcal{B} are polynomial invariants. Indeed, the collection of these polynomials defines a variety of smaller dimension than the one defined by each of them. The resulting speedup can be up to 100 times faster (e.g. in Example 3 “Squares”).

Another observation is that the value $\binom{n+d}{d}$ for M is a rough overestimate that corresponds to the worst case (when there is no polynomial invariant). The choice of such a parameter can be then left to the user depending of the example. In particular, most of the time, as soon as one linear equation $g(F^k(\mathbf{a}), \mathbf{y})$ is a linear combination of the others, then it is the case for all following $g(F^l(\mathbf{a}), \mathbf{y})$, for $k < l \leq \binom{n+d}{d}$.

5.2 Experimental results

In Table 2, we compare our computation timings with those in [1, Table 2]. A notable distinction is that *our approach is global as we compute all polynomial invariants up to a specified degree, whereas theirs is local as they generates only a subset of them*. However, note that the algorithm of [1] can handle probabilistic loops, whereas ours is limited to deterministic ones.

Algorithm 2 excels in cases where the dimension of truncated ideals is small. In certain examples, we use the flexibility of our algorithm to find polynomial invariants : we compute all polynomial invariants with fixed support (the set of non-zero coefficients), instead of the one up to a some degree. In most examples, the validity of these candidates as polynomial invariants is verified. Notably, Gr  bner bases computations is required for each iteration in Algorithm 1 to address the ideal membership problem. However, Table 1 indicates that the number of iterations denoted by N_{\max} within Algorithm 1 is quite small.

Comparing timings in Table 2, our approach outperforms [1] for small degrees, but the latter demonstrates better performance for larger degrees. Importantly, the approach in [1] cannot guarantee that it generates all polynomial invariants up to a given degree, while Algorithm 2 ensures that the computed polynomial invariants form a basis for truncated ideals. Moreover, even when providing the same polynomial invariants as [1], we verify that there are no additional polynomial invariants beyond the computed ones.

Degree	1		2		3		4	
Benchmark	N_{max}	D	N_{max}	D	N_{max}	D	N_{max}	D
Floor function	0	1	2	5	3	13	3	26
Ex 2	0	0	0	1	2	3	2	6
Squares (Ex 3)	0	1	3	5	3	13	3	26
Fib (Ex 4)	0	0	0	0	0	0	0	1
Fib1	0	0	0	0	0	1	2	4
Fib2	0	0	0	0	0	1	TL	TL
Fib3	0	0	0	0	0	1	3	4
Nagata	0	1	2	5	3	13	5	26
Yagzhev9	3	3	3	*6	TL	TL	TL	TL
Yagzhev11	0	0	0	0	0	*1	TL	TL

TL = Timeout (240 seconds); * = Found some invariant of a given degree.

Table 1: Features of the benchmarks. D = the dimension of the corresponding truncated ideal; N_{max} = the largest number of iterations used in Algorithm 1 inside Algorithm 2.

Degree	1		2		3		4	
Benchmark	Ours	UnSol	Ours	UnSol	Ours	UnSol	Ours	UnSol
Floor function	0.014✓	—	0.047✓	—	0.144✓	—	0.42✓	—
Ex 2	0.0097✓	—	0.033✓	—	0.071✓	—	0.19✓	—
Squares	0.0151 ✓	*0.95	0.085 ✓	*1.09	0.25 ✓	*1.66	1.6 ✓	2.72°
Fib	0.008✓	—	0.017✓	—	0.036✓	—	0.082✓	—
Fib1	0.014 ✓	0.31°	0.046 ✓	0.41°	0.17 ✓	*0.91	37.07✓	2.02 °
Fib2	0.017 ✓	0.31°	0.056 ✓	0.49°	12.62✓	*1.34	TL	3.06 °
Fib3	0.013 ✓	0.31°	0.056 ✓	0.48°	0.225 ✓	*1.73	7.11✓	3.46 °
Nagata	0.014 ✓	0.33°	0.057 ✓	*1.04	0.09 ✓	*3.49	0.19 ✓	*8.19
Yagzhev9	0.046 ✓	0.47°	18.69	*2.44✓	TL	*24.31	TL	TL
Yagzhev11	0.075 ✓	0.59°	61.4✓	7.12°	*1.048	*39.27	TL	TL

TL = Timeout (240 seconds); * = Found some invariant of a given degree;

° = Found no invariant; ✓ = Found all invariant up to a given degree

Table 2: The computation times in seconds required to obtain a basis for truncated invariant ideals at their respective degrees.

A EXAMPLES AND BENCHMARKS

Squares

$(x_1, x_2, x_3) = (-1, -1, 1)$
while true do

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \leftarrow \begin{pmatrix} 2x_1 + x_2^2 + x_3 \\ 2x_2 - x_2^2 + 2x_3 \\ 1 - x_3 \end{pmatrix}$$

end while

Fib1

$(x_1, x_2, x_3) = (2, 1, 1)$
while true do

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \leftarrow \begin{pmatrix} x_2 \\ x_3 \\ 2x_2x_3 - x_1 \end{pmatrix}$$

end while

Fib2

$(x_1, x_2, x_3) = (2, -1, 1)$
while true do

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \leftarrow \begin{pmatrix} x_2 \\ 2x_1x_3 - x_2 \\ 4x_1x_2x_3 - 2x_1^2 - 2x_2^2 + 1 \end{pmatrix}$$

end while

Fib3

$(x_1, x_2, x_3) = (3, -2, 1)$
while true do

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \leftarrow \begin{pmatrix} 1 + x_1 + x_2 + x_1x_2 - x_3 \\ x_1 \\ x_2 \end{pmatrix}$$

end while

Floorfunction

$(x_1, x_2, x_3) = (0, 1, 1)$
while $y \leq N$ do

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \leftarrow \begin{pmatrix} x_1 + 1 \\ x_2 + x_3 + 2 \\ x_3 + 2 \end{pmatrix}$$

end while

Nagata

$(x_1, x_2, x_3) = (3, -2, 5)$
while true do

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} \leftarrow \begin{pmatrix} x_1 - 2(x_1x_3 + x_2^2)x_2 - (x_1x_3 + x_2^2)^2x_3 \\ x_2 + (x_1x_3 + x_2^2)x_3 \\ x_3 \end{pmatrix}$$

end while

Yagzhev9

$(x_1, \dots, x_9) = (2, -3, 1, 4, -1, 7, -4, 3, 2)$
while true do

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \\ x_9 \end{pmatrix} \leftarrow \begin{pmatrix} x_1 + x_1x_7x_9 + x_2x_9^2 \\ x_2 - x_1x_7^2 - x_2x_7x_9 \\ x_3 + x_3x_7x_9 + x_4x_9^2 \\ x_4 - x_3x_7^2 - x_4x_7x_9 \\ x_5 + x_5x_7x_9 + x_6x_9^2 \\ x_6 - x_5x_7^2 - x_6x_7x_9 \\ x_7 + (x_1x_4 - x_2x_3)x_9 \\ x_8 + (x_3x_6 - x_4x_5)x_9 \\ (x_9 + (x_1x_4 - x_2x_3)x_8 \\ - (x_3x_6 + x_4x_5)x_7) \end{pmatrix}$$

end while

Yagzhev11

$(x_1, \dots, x_{11}) = (3, -1, 2, 1, -5, -1, 3, 4, -1, 3, 2)$
while true do

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \\ x_9 \\ x_{10} \\ x_{11} \end{pmatrix} \leftarrow \begin{pmatrix} x_1 - x_3x_{10}^2 \\ x_2 - x_3x_{11}^2 \\ x_3 + x_1x_{11}^2 - x_2x_{10}^2 \\ x_4 - x_6x_{10}^2 \\ x_5 - x_6x_{11}^2 \\ x_6 + x_4x_{11}^2 - x_5x_{10}^2 \\ x_7 - x_9x_{10}^2 \\ x_8 - x_9x_{11}^2 \\ x_9 + x_7x_{11}^2 - x_8x_{10}^2 \\ x_{10} - \det(A) \\ x_{11} - x_{10}^3 \end{pmatrix}$$

end while

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