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Coordinate Basis Transformation Based Linearisation Method for Ordinary Differential Equations

Danlei Ye^{1,3}, Xin Jiang^{1,3,*}, Guanying Huo^{1,3}, Cheng Su^{1,3}, Zehong Lu², Bolun Wang^{1,3} and Zhiming Zheng^{1,3}

¹LMIB and School of Mathematics and Systems Science, Beihang University, Beijing, China

²School of Mathematical Science, Peking University, Beijing, China

³Beijing Advanced Innovation Center for Big Data and Brain Computing (BDBC), Beihang University, Beijing, China

*Corresponding author

Abstract — In this paper, a linearisation method for nonlinear ordinary differential equations based on coordinate basis transformation is proposed. In order to transform a nonlinear equation into a linear one, we introduce a general framework to extract the coordinate basis of the differential system. Actually, by introducing a mapping function predefined, this extraction method could connect each state variable of the new system with the variables in the original one. Further, by using vector space for the closure of the Lie derivative, we can analytically obtain the mapping based on the subspace iteration. We argue that this method might be efficient in analyzing and solving ordinary differential systems of complex nonlinear forms.

Keywords- coordinate basis; linearisation; Lie derivative

I. Introduction

Differential equations are generally used in the analysis of complex systems from various fields, including physical dynamics, chemistry, engineering, economics and demographics. It has been developed with calculus, and its theory and method has been one of the key points in the development of modern science and technology. Almost all of the equations that describe the actual problems are non-linear. Along with the need of precise descriptions and solving practical problems, as well as the emergence and improvement of large computer, various problems of solving ordinary differential equation increasingly attract the attention of scientists and engineers. However, the studies of nonlinear equations mostly stay on qualitative analyses and numerical solutions. In 1841, Liouvile proved that the Riccati equation

$$\frac{dx}{dt} = a(t)x^2 + b(t)x + c(t)$$
(1.1)

cannot be solved by the elementary integral method except for some special types. For example, the function $\frac{dx}{dt} = x^2 + y^2$,

which is simple in form, cannot be solved through elementary integration. Until 2010, this problem was solved with multiple integral iterative series methods [15]. In the 19th century, some problems in astrophysics and other technical sciences needed to study the local and global nature of solutions of complex differential equations. Since a large number of differential equations could not be solved by the elementary integral method, researchers discussed numerical solutions, the properties of solutions about the structure and characteristics of the

differential equations or the distribution of curves defined by the differential equations [3-7].

With the vigorous development of nonlinear scientific research, people not only need to discuss individual solutions of differential equations, but also try to understand the general trend and structure of a class of solutions or all the solutions. For example, the number and stability of the solution of nonlinear equations are studied by branching theory. The branch theory also provides a basis for further study of the corresponding dynamic system.

At present, there is no general solution to the nonlinear equation. As mentioned above, nonlinear ordinary differential equations are difficult to calculate and analyze, and the general mathematical method is invalid for nonlinear normal micro systems. Linear ordinary differential equations, in contrast, are more intuitive to show the relationship between each parameter. Moreover, a relatively complete solution and qualitative analysis method have been developed for the given linear ordinary differential equations. Therefore, we can make breakthroughs in the study of nonlinear ordinary differential equation by transforming nonlinear equations into analytically linear equations [1, 2].

Some quasilinearization techniques [8] and linearisation examples that satisfy the local property [9,13,14] was presented and proved. In 1973, Liao Shantao presented the theory of canonical equations [10-12]. The main idea of the theory of canonical equations was to return the vector field of the manifold to the vector field in the Euclidean space. It transformed the properties of the phase diagrams of the ordinary differential system in the manifold transformed into ordinary differential equations in Euclidean space, which was achieved by the activity frame.

The theory of canonical equations is quite complete, but the method is too complicated to carry out. Therefore, we discuss the linearisation problem in another way. In this paper, we focus on the extraction of the coordinate basis, mapping the trajectories of the nonlinear space to the linear space. It is more straightforward of the coordinate basis transformation based linearisation method for ordinary differential equation although it is still in its infancy.

This paper is organized as follows. In Section 2, we introduce some definitions and theorems on differential system and vector space. In Section 3, the linearisation method of coordinate basis



transformation is showed. In Section 4, we give two examples of the linearisation experiment. We summarize our conclusion and give some expectation in the last section.

II. RELATED WORK

In this section we will introduce some definitions and theorems on differential system and vector space, which are prepared for our linearisation method.S

A. Mapping and Similarity on Lipschitz Continuous Vector

Consider a mapping: $\alpha:\mathbb{R}^k\to\mathbb{R}^l$. Given a set $S\subseteq\mathbb{R}^k$, $\alpha(S)$ is defined by applying α to all the elements of S. Similarly, the reflect set is $\alpha^{-1}(T):\{\alpha(s)\in T\}$. Define $\mathscr{F}\langle X_0,\mathscr{F},X_I\rangle$ to be a continuous system on the variate $\vec{x}:(x_1,\ldots,x_n)$ and $\mathscr{F}\langle Y_0,\mathscr{G},Y_I\rangle$ to be a continuous system on the variate $\vec{y}:(y_1,\ldots,y_n)$.

Def.1 (Simulation). \mathscr{T} is thought to simulate \mathscr{T} iff there exists a smooth mapping $\alpha: \mathbb{R}^n \to \mathbb{R}^m$, in which:

- 1. $Y_0 \supseteq X_0$ and $Y_I \supseteq X_I$.
- 2. For any trajectory $\tau:[0,T)\to X_I$ in $\mathscr S$, $\alpha\circ\tau$ is the trajectory in $\mathscr T$.

A simulation relation means that any time trajectory in $\mathscr S$ can be mapped into a trajectory in $\mathscr T$ through α . However, the inversion may not exist since α do not have to be reversible. In other words, there may be some time trajectories in $\mathscr T$ which can not be mapped into by any trajectory in $\mathscr S$.

 $\mathcal F$ and $\mathcal F$ are defined as Lipschitz continuous vector field.

Them.1 \mathcal{T} simulates \mathcal{S} if the following conditions are satisfied:

- 1. $Y_0 \supseteq X_0$.
- 2. $Y_I \supseteq X_I$.
- 3. $\mathscr{G}(\alpha(\vec{x})) = J_{\alpha} \cdot \mathscr{F}(\vec{x})$, where J_{α} is the Jacobian:

$$J_{\alpha}(x_{1},...,x_{n}) = \begin{pmatrix} \frac{\partial \alpha_{1}}{\partial x_{1}} & \cdots & \frac{\partial \alpha_{1}}{\partial x_{n}} \\ \vdots & \ddots & \vdots \\ \frac{\partial \alpha_{m}}{\partial x_{1}} & \cdots & \frac{\partial \alpha_{m}}{\partial x_{n}} \end{pmatrix}$$

and $\alpha(\vec{x}) = (\alpha_1(\vec{x}), ..., \alpha_m(\vec{x})), \alpha : \mathbb{R}^n \to \mathbb{R}$.

Proof. Let τ be a trajectory over \vec{x} of the system \mathscr{S} . Note that at any time $t \in [0,T)$, $\frac{d\tau_x}{dt} = \mathscr{F}(\tau_x(t))$

We want to proof that $\tau_y(t) = \alpha(\tau_x(t))$ is a time trajectory in $\mathscr S$. Since $\tau_x(0) \in X_0$, it can be concluded that

 $\tau_y(0) = \alpha(\tau_x(0)) \in Y_0$. Since $\tau_x(t) \in X_I$ is established for any $t \in [0,T)$, then $\tau_y(t) = \alpha(\tau_x(t)) \in Y_I$. Differentiate τ_y , then we achieve

$$\frac{d\tau_{y}}{dt} = \frac{d\alpha(\tau_{x}(t))}{dt} = J_{\alpha} \cdot \mathscr{F}(\alpha(\tau_{x}(t))) = \mathscr{G}(\tau_{y}(t)). \tag{2.1}$$

Therefor $\tau_y = \alpha \circ \tau_x$ conforms to the dynamical system \mathscr{T} . Through Lipschitz continuity of \mathscr{G} , it shows that τ_y is the unique trajectory starting from $\alpha \circ \tau_x(0)$.

It is important to note that, in normal conditions a trajectory $\tau_y(t) = \alpha \circ \tau_x(t)$ may exist in a longer time interval than [0,T) over which $\tau_y(t)$ is assumed to be defined.

Them.2 Let \mathscr{T} simulate \mathscr{S} through a mapping α . If $Y \supseteq Y_I$ is a positive invariant set of \mathscr{T} , then $\alpha^{-1}(Y) \cap X_I$ is a positive invariant set of \mathscr{S} .

Def.2 (Linearizing Coordinate Bases Transformation). Let $\mathscr S$ be a nonlinear system. α is a linearizing coordinate bases transformation if it maps every trajectory from $\mathscr S$ to a affine system $\mathscr T$. In other words, α insures that $\mathscr S$ simulates a affine system $\mathscr T$.

We search for a mapping α to obtain $\mathcal T$ which simulates a given system $\mathcal F$ through α . In other words, we search for a mapping α that satisfies

$$J_{\alpha}(\vec{x}) \cdot \mathcal{F}(\vec{x}) = A(\vec{x}) + \vec{b}$$
 (2.2)

A,b are constant matrices. After finding out the mapping, we can always find out the appropriate initial and invariance conditions for the simulating system \mathscr{T} whose dynamics will be given by $\mathscr{G}(\vec{y}) = A(\vec{y}) + \vec{b}$.

Def.3 (Lie Derivative). Let $\alpha(\vec{x}) = (\alpha_1(\vec{x}), ..., \alpha_m(\vec{x}))$ be a smooth mapping $\alpha : \mathbb{R}^n \to \mathbb{R}^m$, wherein each $\alpha_i : \mathbb{R}^n \to \mathbb{R}$. $\mathscr{L}_F(\alpha_i(\vec{x})) = (\nabla \alpha_i) \cdot \mathscr{F}(\vec{x})$ indicates the Lie derivative of the function $\alpha_i(\vec{x})$ in the vector field \mathscr{F} .

Lemma.1

$$J_{\alpha}(\vec{x}) \cdot \mathcal{F}(\vec{x}) = A(\vec{x}) + \vec{b}$$

Proof. Since Jacobian J_{α} :

$$J_{\alpha}(x_{1},...x_{n}) = \begin{pmatrix} \frac{\partial y_{1}}{\partial x_{1}} & ... & \frac{\partial y_{1}}{\partial x_{n}} \\ \vdots & \ddots & \vdots \\ \frac{\partial y_{m}}{\partial x_{1}} & ... & \frac{\partial y_{m}}{\partial x_{n}} \end{pmatrix} = \begin{pmatrix} \nabla \alpha_{1} \\ \vdots \\ \nabla \alpha_{n} \end{pmatrix}$$
(2.3)



It can be obtained that

$$J_{\alpha} \cdot \mathscr{F} = \begin{pmatrix} (\nabla \alpha_{1}) \cdot \mathscr{F} \\ \vdots \\ (\nabla \alpha_{n}) \cdot \mathscr{F} \end{pmatrix} = \begin{pmatrix} \mathscr{L}_{F} (\alpha_{1}(\vec{x})) \\ \vdots \\ \mathscr{L}_{F} (\alpha_{m}(\vec{x})) \end{pmatrix}$$
(2.4)

Given functions $\alpha_1, \alpha_2, ..., \alpha_m : \mathbb{R}^n \to \mathbb{R}$ and constant equations $1: \mathbb{R}^n \to \mathbb{R}$, we consider the vector space generated by these equations:

$$Span(1,\alpha_1,\ldots,\alpha_m) = \left\{ c_0 \cdot 1 + \sum_{i=1}^m c_i \alpha_i \, \middle| \, c_0,\ldots,c_m \in \mathbb{R} \right\}$$
(2.5)

B. The Closure of Vector Space

Them.3 (Vector Space Closure Theorem). The vector space is closed under Lie derivative. In other words, a mapping $\alpha:(\alpha_1,\ldots,\alpha_m)$ represents a linearizing coordinate bases transformation of a system iff the vector space $V:Span(1,\alpha_1,\ldots,\alpha_m)$ is closure under the calculation of Lie derivative.

For example, $\forall g \in V, \mathcal{L}_F(g) \in V$.

Proof. Let α be a linearizing coordinate bases transformation which maps trajectories in $\mathscr F$ into trajectories in $\mathscr F$:

$$\frac{d\vec{y}}{dt} = A\vec{y} + \vec{b} \tag{2.6}$$

Therefor to each α_i ,

$$\mathcal{L}_{F}(\alpha_{i}) = b_{i} + \sum_{j} A_{ij} \alpha_{j}$$
(2.7)

Any element $\beta \in V$ can be written as $\beta = c_0 + \sum_k c_k \alpha_k$ which is a linear combination of the coordinate bases of the vector space. Using (2.6) we can obtain $\mathscr{L}_F(\beta)$ which is a linear combination of α_i and 1. Therefor it can be proofed that $\mathscr{L}_F(\beta) \in V$.

On the contrary, if V is closure under the calculate of Lie derivative, then its bases $1, \alpha_1, \dots, \alpha_m$ satisfy the condition $\mathcal{L}_F(\alpha_i) = b_i 1 + \sum_i a_{ij} \alpha_j$.

From Lemma1,

$$J_{\alpha} \cdot \mathcal{F} = \begin{pmatrix} \mathcal{L}_{F} \left(\alpha_{1}(\vec{x}) \right) \\ \vdots \\ \mathcal{L}_{F} \left(\alpha_{m}(\vec{x}) \right) \end{pmatrix} = A\alpha(\vec{x}) + \vec{b}$$

$$(2.8)$$

wherein $A = [\alpha_{ij}], \vec{b} = (b_i)$. All above results demonstrate that α is a linearizing coordinate bases transformation.

III. COORDINATE BASIS TRANSFORMATION

In this section, we will present the linearisation method of coordinate basis transformation for ordinary differential equation space.

A. Subspace Iterative Method

If the linearising method of coordinate bases transformation exists, it could be found out by the closure of vector space under Lie derivative. Our goal is to find out a vector space that consists of some functions, while this space is closed under the action of Lie derivative.

Given a set of functions $B = \{f_1, ..., f_k\}$, we use Span(B) to represent the vector established by the functions in B.

$$Span(B) = \left\{ \sum_{j=1}^{k} a_j f_j(\vec{x}) \middle| a_j \in \mathbb{R} \right\}$$
(3.1)

We use a method of subspace iteration as follows:

- 1. Choose an initial vector space $V_0 = Span(\{\alpha_1, ..., \alpha_N\})$.
- 2. Use V_i to define V_{i+1} at each step, wherein V_{i+1} is the subspace of V_i .
- 3. Stop the operation when $V_{n+1} = V_n$. IF V_n is a non-trivial vector space, then we can achieve a non-trivial linear coordinate basis transformation from the basis of V_n .

At step 1, we consider to choose the initial vector space V_0 . To those ordinary differential equations with polynomial on the right-hand side, usually V_0 can be made up of all the monomial expressions whose degrees are less than a given value. Each function f_i in V_0 can be written as $\sum c_i \alpha_i$, wherein c_i is a

parameter. V_0 can also be represented by these functions in it: $V_0 = Span(\{f_0, ..., f_k\})$. There are two methods to establish the initial space: one is setting a high degree of the initial bases of V_0 and then the degrees would reduce with the bases reduced during the iteration. The other one is setting the same degree N of the initial bases as the highest degree of the polynomial on the right-hand side of the ordinary differential equation, and if the initial space was proved to be improper after the iterative computations, then we would reset the initial degree as N+1. In some particular situations, the initial space can be fixed. It will be mentioned later in this paper.

At step 2, V_{i+1} is calculated as follows:

$$V_{i+1} = \mathcal{D}(V_i) = \left\{ f \in V_i \middle| \mathcal{L}_F(f) \in V_i \right\}$$
(3.2)



Lemma.2 $V_{i+1} \subseteq \mathcal{D}(V_i)$ is the subspace of V_i , if V_i is a vector space.

Proof. Suppose that V_i is made up of the bases $(1,\alpha_1,\ldots,\alpha_k)$, then it is obvious that $V_{i+1}\subseteq \mathscr{D}(V_i)$. Now we want to prove that V_{i+1} is a vector space. First $1\in V_{i+1}$. Let f_1,\ldots,f_l are functions in V_{i+1} , it can be obtained that $\mathscr{D}(f_1),\ldots,\mathscr{D}(f_l)\in V_i$. The affine combination of the functions is $f:a_01+\sum_{j=1}^l a_jf_j$, wherein a_1,\ldots,a_l . Its Lie derivative is

$$\mathcal{L}_{F} f\left(a_{0} 1 + \sum_{j=1}^{l} a_{j} f_{j}\right) = a_{j} \sum_{j=1}^{l} \mathcal{L}_{F} \left(f\right)_{j}$$

$$(3.3)$$

Therefor $\mathscr{L}_F f$ can be written as the linear combination of the Lie derivatives of f_1,\ldots,f_l in V_i . Thus $\mathscr{L}_F (f) \in V$ and then $f \in V_{i+1}$. Each linear combination of the elements in V_{i+1} belongs to V_i as well.

Them.4 Given an initial vector space V_0 and vector field \mathscr{F} , the space will be converged to a subspace $V^* \subseteq V_0$ through limited steps of iteration. Let $\alpha_1, \ldots, \alpha_m$ be the base functions which establish V^* ,

- 1. The transformation $\alpha:(\alpha_1,...,\alpha_m)$, which is made up of the base function of the space that we obtain finally, is linear.
- 2. To each linearising coordinate bases transformation $\beta:(\beta_1,...,\beta_m), \beta \in V^*$ for each $\beta \in V_0$.

Proof. If $V_{i+1} \subseteq V_i$ is converged through the iteration, then the degree of V_{i+1} would be less than V_i . Since the degree of V_0 is limited, the upper limit of the iteration is the same as the number of the base equations of V_0 .

We suppose that there exists a linearisation coordinate basis transformation $\beta:(\beta_1,\ldots,\beta_m)$ which satisfied $\beta_i\in V_0$. Note that the space U established by $1,\beta_1,\ldots,\beta_m$ is the subset of V_0 . It also can be proved that if $U\subseteq V_i$, then $U\subseteq V_{i+1}$. Through the inductive method, $U\subseteq V^*$ is proved.

B. The Algorithm

Suppose the initial ODE: $\dot{X} = F(X)$, wherein $X = (x_1, ..., x_n)^T$, $F = (f_1, ..., f_n)^T$.

1. Observe the the right-hand side of the equation to get the highest degree N. Set non-negative integer parameters l, m, l_m, i, k .

2. Establish a set

$$V = \left\{ \prod_{\substack{1 \le m \le n \\ \sum l_m = l}} x_m^{l_m} \left| 0 \le l \le N \right. \right\} \triangleq \left\{ y_k \left| 1 \le k \le M \right. \right\}.$$

- 3. Set $f = \sum_{k \in \tau_v} c_k y_k$, wherein τ_k is the set of the indexes of
 - *V*. Calculate the Lie derivative $\mathscr{L}_F(f) = \sum_{i=1}^n \left(\frac{\partial f}{\partial x_i}\right) \cdot f_i$.
- 4. Compare the Lie derivative $\mathscr{L}_F(f)$ and the right-hand side of the equation F(X). If an element y_k did not exist in F(X) or V, let the coefficient of y_k in the integrated $\mathscr{L}_F(f)$ be zero.
- 5. Arrange $\mathcal{L}_F(f)$ once again. Select the nonzero terms to establish a new set V.
- 6. If $V_1 = \Phi$, then we set N = N + 1 and return to step 2; if $V_1 = V$, then stop the steps and output V; otherwise, set $V = V_1$ and return to step 3.

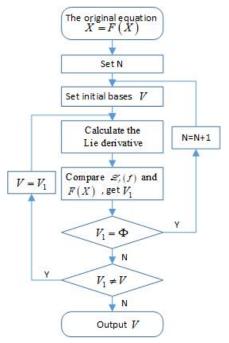


FIGURE I. ALGORITHM FLOW CHART OF THE SUBSPACE ITERATIVE METHOD

C. The Initial Bases Selection

It seems to be straightforward to linearising any non-linear ordinary differential equation while the generic algorithm has been applied. Nevertheless, because of the uncertainty of the initial base selection, the iterative algorithm will be quite tedious when the degree of the initial differential system is high. And in the actual engineering problems, most often we just need to analyze systems of ordinary differential equations with three or two elements.



We found that for a system of three-dimensional ordinary differential equations, when the right-hand side of the equation only includes constant terms, one degree terms and cross terms, it is able to set

$$\left(x_1^2, x_1 x_2, x_1 x_3, x_1, x_2^2, x_2 x_3, x_2, x_3^2, x_3, 1\right) \tag{3.4}$$

as the initial bases. In other words, any three-dimensional ordinary differential equations which only includes constant terms, one degree terms and cross terms in the right-hand side of the equations, can be transformed into linear ordinary differential equations with no more than nine elements. More experiments are shown in the next section.

IV. EXPERIMENT

In this section, we present two examples of the linearisation method based on coordinate basis transformation.

Exp.1 Given a non-linear ordinary differential equation

$$\begin{cases} \dot{x}_1 = 9x_1 + 3x_2 + 10x_3 + 10x_1x_2 + x_1x_3 + 8x_2x_3 + 8 \\ \dot{x}_2 = 6x_1 + 5x_2 + 2x_3 + 5x_1x_2 + 4x_1x_3 + 10x_2x_3 + 9 \\ \dot{x}_3 = x_1 + 10x_2 + 10x_3 + 8x_1x_2 + 9x_1x_3 + 7x_2x_3 + 1 \end{cases}$$

$$(4.1)$$

Based on the previous conclusion, we directly set the initial bases $Y = (y_1, ..., y_q, 1)$:

$$\begin{cases} y_1 = x_1^2 & y_2 = x_1 x_2 & y_3 = x_1 x_3 \\ y_4 = x_1 & y_5 = x_2^2 & y_6 = x_2 x_3 \\ y_7 = x_2 & y_8 = x_3^2 & y_9 = x_3 \end{cases}$$

$$(4.2)$$

Then the linearising result shows as follows:

$$\begin{cases} \dot{y}_1 = 18y_1 + 6y_2 + 20y_3 + 16y_4 \\ \dot{y}_2 = 6y_1 + 14y_2 + 2y_3 + 9y_4 + 3y_5 + 10y_6 + 8y_7 \\ \dot{y}_3 = y_1 + 10y_2 + 19y_3 + y_4 + 3y_6 + 10y_8 + 8y_9 \\ \dot{y}_4 = 10y_2 + y_3 + 9y_4 + 8y_6 + 3y_7 + 10y_9 + 8 \\ \dot{y}_5 = 12y_2 + 10y_5 + 4y_6 + 18y_7 \\ \dot{y}_6 = y_2 + 6y_3 + 10y_5 + 15y_6 + y_7 + 2y_8 + 9y_9 \\ \dot{y}_7 = 5y_2 + 4y_3 + 6y_4 + 10y_6 + 5y_7 + 2y_9 + 9 \\ \dot{y}_8 = 2y_3 + 20y_6 + 20y_8 + 2y_9 \\ \dot{y}_9 = 8y_2 + 9y_3 + y_4 + 7y_6 + 10y_7 + 10y_9 + 1 \end{cases}$$

$$(4.3)$$

The original three-dimensional nonlinear ordinary differential equation is converted to a nine-dimensional linear ordinary differential equation. The other experiment will present that the linearising result of the same kind of nonlinear ordinary differential equation may be more concise.

Exp.2 Given a non-linear ordinary differential equation

$$\begin{cases} \dot{x}_1 = 3x_1 + x_2 + 4x_3 + 4x_1x_2 + 7x_1x_3 + 7x_2x_3 + 4\\ \dot{x}_2 = 10x_1 + 7x_2 + 5x_3 + 10x_1x_2 + 7x_1x_3 + 7x_2x_3 + 2\\ \dot{x}_3 = 9x_1 + 3x_2 + 9x_3 + 3x_1x_2 + 5x_1x_3 + 2x_2x_3 + 5 \end{cases}$$
(4.4)

We select the same initial bases as in Exp 1 and after the iterative steps, the bases have been simplified:

$$\begin{cases} y_1 = x_1 x_2 & y_2 = x_1 x_3 & y_3 = x_2 x_3 \\ y_4 = x_1 & y_5 = x_2 & y_6 = x_3 \end{cases}$$
 (4.5)

Then the linearising result is:

$$\begin{cases} y_1 = 10y_1 + 5y_2 + 2y_3 + 4y_4 + 4y_5 \\ y_2 = 3y_1 + 12y_2 + 5y_3 + y_4 + 4y_6 \\ y_3 = 4y_1 + 7y_2 + 3y_3 + 7y_4 + y_5 + 4y_6 + 4 \\ y_4 = 9y_1 + 10y_2 + 16y_4 + 5y_5 + 2y_6 \\ y_5 = 10y_1 + 7y_2 + 10y_3 + 7y_4 + 7y_5 + 5y_6 + 2 \\ y_6 = 3y_1 + 5y_2 + 9y_3 + 2y_4 + 3y_5 + 9y_6 + 5 \end{cases}$$

$$(4.6)$$

V. CONCLUSION AND EXPECTATION

In this paper, we demonstrate that in most cases there exist a linearising coordinate bases transformation which could map a given non-linear ordinary differential system into a linear one. It is shown that the coordinate bases can be obtained through limited iteration steps. It is interesting to find that if a three-dimensional ordinary differential system only includes constant terms, one degree terms and cross terms, it can be always transformed into a linear ordinary differential system with no more than nine term. We argue that this linearising algorithm is exercisable in case of given appropriate initial bases. However, searching for the affine transformation of multivariate high-dimensional non-linear ordinary differential equations is always tedious. We might investigate this problem in our future studies.

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