

Research Article

A New Algorithm to Approximate Bivariate Matrix Function via Newton-Thiele Type Formula

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A new method for computing the approximation of bivariate matrix function is introduced. It uses the construction of bivariate Newton-Thiele type matrix rational interpolants on a rectangular grid. The rational interpolant is of the form motivated by Tan and Fang (2000), which is combined by Newton interpolant and branched continued fractions, with scalar denominator. The matrix quotients are based on the generalized inverse for a matrix which is introduced by C. Gu the author of this paper, and it is effective in continued fraction interpolation. The algorithm and some other important conclusions such as divisibility and characterization are given. In the end, two examples are also given to show the effectiveness of the algorithm. The numerical results of the second example show that the algorithm of this paper is better than the method of Thieletype matrix-valued rational interpolant in Gu (1997).

1. Introduction

Matrix-valued rational interpolation and approximation theory have further practical application in many fields, such as in automatic control theory, computer science, and elementary particle physics [1]. Kuchminska et al. proved an analog of the van Vlerk theorem and constructed an interpolation formula of the Newton-Thiele type in [2]. Tan and Fang in [3] put emphasis on the study of Newton-Thiele bivariate rational interpolants. Graves-Morris provided a practical Thiele-fraction method for rational interpolation of vectors, based on the Samelson inverse in [4]. Gu et al. generalized the definition of Samelson inverse to the case of matrices and applied it to deal with the problems of rational interpolation of matrices [5–9]. Bose and Basu gave the existence, nonuniqueness, and recursive computation of the two-dimensional matrix Padé approximants in [10].

In this paper, we introduce a bivariate matrix-valued rational interpolant, with scalar denominator, in Newton-Thiele form motivated by [3]. The construction of the interpolant is combined by the classic methods: Newton interpolant and branched continued fractions. As we know, branched continued fraction has been studied by Cuyt and

Verdonk and Siemaszko [11, 12] and many other authors. The interpolant of this paper is based on using the generalized inverse of matrices. A new algorithm to approximate bivariate matrix function is given and some examples are also put to show the effectiveness of the algorithm which is much better than the method mentioned in [9].

First, we will give the definition of the so-called generalized inverse of a matrix. Let $C^{u \times v}$ consists of all $u \times v$ matrices with their elements in the complex plan C , and let $A = (a_{ij})$, $B = (b_{ij})$, and $A, B \in C^{u \times v}$.

Definition 1 (see [8]). The scalar product of matrices A and B is defined by

$$A \cdot B = \sum_{i=1}^u \sum_{j=1}^v a_{ij} b_{ij} = \text{tr}(AB^*), \quad (1)$$

where B^* denotes the transpose of B and the Euclidean norm of A is given by

$$\|A\| = \left(\sum_{i=1}^u \sum_{j=1}^v |a_{ij}|^2 \right)^{1/2}. \quad (2)$$

It follows from Definition 1 and (2) that

$$A \cdot \bar{A} = \sum_{i=1}^u \sum_{j=1}^v |a_{ij}|^2 = \|A\|^2 = \text{tr}(A\bar{A}^*), \quad (3)$$

where \bar{A} denotes the complex conjugate matrix of A and \bar{A}^* denotes the complex conjugate transpose matrix of A . On the basis of (2) and (3), the generalized inverse of the matrix A is defined as

$$A_r^{-1} = \frac{1}{A} = \frac{\bar{A}}{\|A\|^2}, \quad A \neq 0, \quad A \in C^{u \times v}, \quad (4)$$

in particular, for $A \in R^{u \times v}$, $A_r^{-1} = 1/A = A/\|A\|^2$, $A \neq 0$.

By means of generalized inverse A_r^{-1} for matrices, we want to define bivariate Newton-Thiele rational interpolants and give the algorithm and some properties on a rectangular grid.

2. Newton-Thiele Interpolation Formula

Let $\Lambda_{m,n} = \{(x_l, y_k) : l = 0, 1, \dots, m, k = 0, 1, \dots, n, (x_l, y_k) \in \mathbb{R}^2\}$.

A matrix-valued set

$$\Pi^{m,n} = \{Y_{l,k} : Y_{l,k} = Y(x_l, y_k) \in C^{u \times v}, (x_l, y_k) \in \Lambda_{m,n}\}. \quad (5)$$

We need to find a bivariate matrix rational function

$$R_{m,n}(x, y) = \frac{N(x, y)}{D(x, y)}, \quad (6)$$

whose numerator $N(x, y)$ is a complex or real polynomial matrix and denominator $D(x, y)$ is a real polynomial, and

$$R_{m,n}(x_l, y_k) = \frac{N(x_l, y_k)}{D(x_l, y_k)} = Y_{l,k}, \quad (x_l, y_k) \in \Lambda_{m,n}. \quad (7)$$

First, some notations need to be given as follows:

$$\varphi_{0,0}(x_l, y_k) = Y_{l,k}, \quad \forall (x_l, y_k) \in \Lambda_{m,n},$$

$$\varphi_{1,0}(x_i, x_j, y_k) = \frac{\varphi_{0,0}(x_j, y_k) - \varphi_{0,0}(x_i, y_k)}{x_j - x_i},$$

$$\begin{aligned} \varphi_{l,0}(x_p, \dots, x_q, x_i, x_j, y_k) \\ = (\varphi_{l-1,0}(x_p, \dots, x_q, x_j, y_k) \\ - \varphi_{l-1,0}(x_p, \dots, x_q, x_i, y_k)) \times (x_j - x_i)^{-1}, \end{aligned}$$

$$\begin{aligned} \varphi_{l,1}(x_p, \dots, x_q, x_i, x_j, y_r, y_s) \\ = (y_s - y_r) \times (\varphi_{l,0}(x_p, \dots, x_q, x_i, x_j, y_s) \\ - \varphi_{l,0}(x_p, \dots, x_q, x_i, x_j, y_r))^{-1}, \\ \varphi_{l,k}(x_p, \dots, x_q, x_i, x_j, y_a, \dots, y_b, y_r, y_s) \\ = (y_s - y_r) \\ \times (\varphi_{l,k-1}(x_p, \dots, x_q, x_i, x_j, y_a, \dots, y_b, y_s) \\ - \varphi_{l,k-1}(x_p, \dots, x_q, x_i, x_j, y_a, \dots, y_b, y_r))^{-1}, \end{aligned} \quad (8)$$

where the first subscript l of φ means the number of nodes $x_p, \dots, x_q, x_i, x_j$ minus 1, and the second subscript k of φ means the number of nodes $y_a, \dots, y_b, y_r, y_s$ minus 1.

For simplicity, we let the nodes $x_p, \dots, x_q, x_i, x_j$ be replaced by x_0, \dots, x_l , and the nodes $y_a, \dots, y_b, y_r, y_s$ be replaced by y_0, \dots, y_k , then we can get the following definition.

Definition 2. By means of generalized inverse (4), we define bivariate Newton-Thiele type matrix blending differences as follows:

$$\begin{aligned} \varphi_{0,0}(x_l, y_k) &= Y_{l,k}, \\ \varphi_{1,0}(x_0, x_1, y_k) &= \frac{\varphi_{0,0}(x_1, y_k) - \varphi_{0,0}(x_0, y_k)}{x_1 - x_0}, \\ \varphi_{l,0}(x_0, \dots, x_l, y_k) \\ &= (\varphi_{l-1,0}(x_0, \dots, x_{l-2}, x_l, y_k) \\ &\quad - \varphi_{l-1,0}(x_0, \dots, x_{l-2}, x_{l-1}, y_k)) \times (x_l - x_{l-1})^{-1}, \\ \varphi_{l,1}(x_0, \dots, x_l, y_0, y_1) \\ &= \frac{(y_1 - y_0)}{\varphi_{l,0}(x_0, \dots, x_l, y_1) - \varphi_{l,0}(x_0, \dots, x_l, y_0)}, \\ \varphi_{0,k}(x_l, y_0, \dots, y_k) \\ &= (y_k - y_{k-1}) \\ &\quad \times (\varphi_{0,k-1}(x_l, y_0, \dots, y_{k-2}, y_k) \\ &\quad - \varphi_{0,k-1}(x_l, y_0, \dots, y_{k-2}, y_{k-1}))^{-1}, \\ \varphi_{l,k}(x_0, x_1, \dots, x_l, y_0, \dots, y_k) \\ &= (y_k - y_{k-1}) \\ &\quad \times (\varphi_{l,k-1}(x_0, \dots, x_l, y_0, \dots, y_{k-2}, y_k) \\ &\quad - \varphi_{l,k-1}(x_0, \dots, x_l, y_0, \dots, y_{k-2}, y_{k-1}))^{-1}. \end{aligned} \quad (9)$$

We assume that for all $l, x_l \neq x_{l-1}$ and

$$\begin{aligned} &\varphi_{l,k-1}(x_0, \dots, x_l, y_0, \dots, y_{k-2}, y_k) \\ &\neq \varphi_{l,k-1}(x_0, \dots, x_l, y_0, \dots, y_{k-2}, y_{k-1}), \quad \forall l, k. \end{aligned} \tag{10}$$

From (9) matrix-valued Newton-Thiele type continued fractions for two-variable function can be constructed as follows:

$$\begin{aligned} R_{m,n}(x, y) &= U_0(y) + U_1(y)(x - x_0) + \dots + U_m(y)(x - x_0) \\ &\times (x - x_1) \dots (x - x_{m-1}), \end{aligned} \tag{11}$$

where for $l = 0, 1, \dots, m$

$$\begin{aligned} U_l(y) &= \varphi_{l,0}(x_0, \dots, x_l, y_0) + \frac{y - y_0}{\varphi_{l,1}(x_0, \dots, x_l, y_0, y_1)} \\ &+ \dots + \frac{y - y_{n-1}}{\varphi_{l,n}(x_0, \dots, x_l, y_0, \dots, y_n)}. \end{aligned} \tag{12}$$

Remark 3. Because of the construction, $D(x, y)$ in $R_{m,n}(x, y)$ mentioned as (6) is actually $D(y)$ and each $U_l(y)$ depends on n .

We can also construct the antithetical form of bivariate matrix continued fraction in light of Definition 4.

Definition 4. By means of generalized inverse (4), we define bivariate antithetical Newton-Thiele type matrix blending differences

$$\begin{aligned} \psi_{0,0}(x_l, y_k) &= Y_{l,k}, \\ \psi_{0,1}(x_l, y_0, y_1) &= \frac{(\psi_{0,0}(x_l, y_1) - \psi_{0,0}(x_l, y_0))}{(y_1 - y_0)} \\ \psi_{0,k}(x_l, y_0, \dots, y_k) &= (\psi_{0,k-1}(x_l, y_0, \dots, y_{k-2}, y_k) \\ &\quad - \psi_{0,k-1}(x_l, y_0, \dots, y_{k-2}, y_{k-1})) \\ &\quad \times (y_k - y_{k-1})^{-1}, \\ \psi_{1,k}(x_0, x_1, y_0, \dots, y_k) &= \frac{(x_1 - x_0)}{\psi_{0,k}(x_1, y_0, \dots, y_k) - \psi_{0,k}(x_0, y_0, \dots, y_k)}, \\ \psi_{l,0}(x_0, \dots, x_l, y_k) &= (x_l - x_{l-1}) \\ &\quad \times (\psi_{l-1,0}(x_0, \dots, x_{l-2}, x_l, y_k) \\ &\quad - \psi_{l-1,0}(x_0, \dots, x_{l-1}, y_k))^{-1}, \end{aligned}$$

$$\begin{aligned} &\psi_{l,k}(x_0, x_1, \dots, x_l, y_0, \dots, y_k) \\ &= (x_l - x_{l-1}) \\ &\quad \times (\psi_{l-1,k}(x_0, \dots, x_{l-2}, x_l, y_0, \dots, y_k) \\ &\quad - \psi_{l-1,k}(x_0, \dots, x_{l-1}, y_0, \dots, y_k))^{-1}. \end{aligned} \tag{13}$$

From (13) we can get the antithetical formula for two-variable function

$$\begin{aligned} \widehat{R}_{m,n}(x, y) &= V_0(x) + V_1(x)(y - y_0) \\ &\quad + \dots + V_n(x)(y - y_0)(y - y_1) \\ &\quad \dots (y - y_{n-1}), \end{aligned} \tag{14}$$

where for $k = 0, 1, \dots, n$

$$\begin{aligned} V_k(x) &= \psi_{0,k}(x_0, y_0, \dots, y_k) \\ &\quad + \frac{x - x_0}{\psi_{1,k}(x_0, x_1, y_0, \dots, y_k)} \\ &\quad + \dots + \frac{x - x_{m-1}}{\psi_{m,k}(x_0, \dots, x_m, y_0, \dots, y_k)}. \end{aligned} \tag{15}$$

Now we will give two theorems about $R_{m,n}(x, y)$ as in (11) and (12) and $\widehat{R}_{m,n}(x, y)$ as in (14) and (15). First of all, some notations need to be defined.

In (12), for $l = 0, 1, \dots, m$, let

$$\begin{aligned} U_l^{(s)}(y) &= \varphi_{l,s}(x_0, x_1, \dots, x_l, y_0, \dots, y_s) \\ &\quad + \frac{y - y_s}{U_l^{(s+1)}(y)}, \quad s = 0, 1, \dots, n - 1, \end{aligned} \tag{16}$$

where

$$\begin{aligned} U_l^{(n)}(y) &= \varphi_{l,n}(x_0, x_1, \dots, x_l, y_0, \dots, y_n), \\ U_l^{(0)}(y) &= U_l(y). \end{aligned} \tag{17}$$

Similarly, in (15), for $k = 0, 1, \dots, n$, let

$$\begin{aligned} V_k^{(s)}(x) &= \psi_{s,k}(x_0, x_1, \dots, x_s, y_0, \dots, y_k) \\ &\quad + \frac{x - x_s}{V_k^{(s+1)}(x)}, \quad s = 0, 1, \dots, m - 1, \end{aligned} \tag{18}$$

where

$$\begin{aligned} V_k^{(m)}(x) &= \psi_{m,k}(x_0, x_1, \dots, x_m, y_0, \dots, y_k), \\ V_k^{(0)}(x) &= V_k(x). \end{aligned} \tag{19}$$

Theorem 5. Let

- (i) $\varphi_{l,k}(x_0, \dots, x_l, y_0, \dots, y_k), l = 0, 1, \dots, m; k = 0, 1, \dots, n$ exist and be nonzero (except for $\varphi_{l,0}(x_0, \dots, x_l, y_0)$),

(ii) $U_l^{(s)}(y) = \varphi_{l,s}(x_0, x_1, \dots, x_l, y_0, \dots, y_s) + ((y - y_s)/U_l^{(s+1)}(y))$ satisfy $U_l^{(s+1)}(y_s) \neq 0, s = 0, 1, \dots, n-1$, then $R_{m,n}(x, y)$ as in (11) and (12) exists such that $R_{m,n}(x_l, y_k) = Y_{lk}, (x_l, y_k) \in \Lambda_{m,n}$.

if the conditions hold, thus for $l = 0, 1, \dots, m; k = 0, 1, \dots, n$

$$U_l(y_k) = U_l^{(0)}(y_k) = \varphi_{l,0} + \frac{y_k - y_0}{\varphi_{l,1}} + \dots + \frac{y_k - y_{k-1}}{|\varphi_{l,n}|} + \frac{y_k - y_k}{|U_l^{(k+1)}(y_k)|} \tag{21}$$

Proof. For simpleness, let

$$\varphi_{l,k}(x_0, \dots, x_l, y_0, \dots, y_k) = \varphi_{l,k} \tag{20}$$

$(y_k - y_k)/U_l^{(k+1)}(y_k) = 0$, since $U_l^{(k+1)}(y_k) \neq 0$, then we get \square

$$\begin{aligned} U_l(y_k) &= \varphi_{l,0}(x_0, \dots, x_l, y_0) + \frac{y_k - y_0}{\varphi_{l,1}(x_0, \dots, x_l, y_0, y_1)} + \dots + \frac{y_k - y_{k-2}}{|\varphi_{l,k-1}(x_0, \dots, x_l, y_0, \dots, y_{k-1})|} \\ &\quad + \frac{y_k - y_{k-1}}{|\varphi_{l,k}(x_0, \dots, x_l, y_0, \dots, y_k)|} \\ &= \varphi_{l,0}(x_0, \dots, x_l, y_0) + \frac{y_k - y_0}{\varphi_{l,1}(x_0, \dots, x_l, y_0, y_1)} + \dots \\ &\quad + \frac{y_k - y_{k-2}}{|\varphi_{l,k-1}(x_0, \dots, x_l, y_0, \dots, y_{k-1})|} \\ &\quad + \frac{y_k - y_{k-1}}{|(y_k - y_{k-1}) \times (\varphi_{l,k-1}(x_0, \dots, x_l, y_0, \dots, y_{k-2}, y_k) - \varphi_{l,k-1}(x_0, \dots, x_l, y_0, \dots, y_{k-1}))^{-1}} \\ &= \varphi_{l,0}(x_0, \dots, x_l, y_0) + \frac{y_k - y_0}{\varphi_{l,1}(x_0, \dots, x_l, y_0, y_1)} + \dots + \frac{y_k - y_{k-2}}{|\varphi_{l,k-1}(x_0, \dots, x_l, y_0, \dots, y_{k-2}, y_k)|} \\ &= \dots = \varphi_{l,0}(x_0, \dots, x_l, y_0) + \frac{y_k - y_0}{\varphi_{l,1}(x_0, \dots, x_l, y_0, y_k)} \\ &= \varphi_{l,0}(x_0, \dots, x_l, y_k). \end{aligned} \tag{22}$$

We can use (9), (11), and (22) to find that

$$\begin{aligned} R_{m,n}(x_l, y_k) &= U_0(y_k) + U_1(y_k)(x_l - x_0) \\ &\quad + \dots + U_l(y_k)(x_l - x_0)(x_l - x_1) \dots (x_l - x_{l-1}) \\ &= Y(x_l, y_k). \end{aligned} \tag{23}$$

We can similarly prove the following result.

Theorem 6. Let

- (i) $\psi_{l,k}(x_0, \dots, x_l, y_0, \dots, y_k), l = 0, 1, \dots, m; k = 0, 1, \dots, n$ exist and be nonzero (except for $\psi_{0,k}(x_0, y_0, \dots, y_k)$),
- (ii) $V_k^{(s)}(x) = \psi_{s,k}(x_0, x_1, \dots, x_s, y_0, \dots, y_k) + ((x - x_s)/V_k^{(s+1)}(x))$ satisfy $V_k^{(s+1)}(x_s) \neq 0, s = 0, 1, \dots, m-1$, then $\widehat{R}_{m,n}(x, y)$ as in (14) and (15) exists such that $\widehat{R}_{m,n}(x_l, y_k) = Y_{lk}, (x_l, y_k) \in \Lambda_{m,n}$.

Now we give the Newton-Thiele Matrix Interpolation Algorithm (NTMIA).

Algorithm 7 (NTMIA). **Input:** $\{(x_i, y_j), Y_{i,j}\} (i = 0, 1, \dots, m; j = 0, 1, \dots, n)$

Output: $\widehat{R}_m(x, y)$.

Step 1. For all $(x_i, y_j) \in \Lambda_{m,n}$, let $\varphi(x_i, y_j) = Y_{i,j}$.

Step 2. For $j = 0, 1, \dots, n; p = 1, 2, \dots, m; i = p, p + 1, \dots, m$, compute

$$\begin{aligned} \varphi_{p,0}(x_0, \dots, x_{p-1}, x_i, y_j) &= \frac{\varphi_{p-1,0}(x_0, \dots, x_{p-2}, x_i, y_j) - \varphi_{p-1,0}(x_0, \dots, x_{p-1}, y_j)}{x_i - x_{p-1}}. \end{aligned} \tag{24}$$

Step 3. For $i = 0, 1, \dots, m$; $q = 1, 2, \dots, n$; $j = q, q + 1, \dots, n$, compute

$$\begin{aligned} & \varphi_{i,q}(x_0, \dots, x_i, y_0, \dots, y_{q-1}, y_j) \\ &= (y_j - y_{q-1}) \times \left(\varphi_{i,q-1}(x_0, \dots, x_i, y_0, \dots, y_{q-2}, y_j) \right. \\ & \quad \left. - \varphi_{i,q-1}(x_0, \dots, x_i, y_0, \dots, y_{q-1}) \right)^{-1} \\ &= \left((y_j - y_{q-1}) \overline{\left(\varphi_{i,q-1}(x_0, \dots, x_i, y_0, \dots, y_{q-2}, y_j) \right)} \right. \\ & \quad \left. - \overline{\varphi_{i,q-1}(x_0, \dots, x_i, y_0, \dots, y_{q-1})} \right) \\ & \quad \times \left(\left\| \varphi(x_0, \dots, x_i, y_0, \dots, y_{q-2}, y_j) \right. \right. \\ & \quad \left. \left. - \varphi(x_0, \dots, x_i, y_0, \dots, y_{q-1}) \right\|^2 \right)^{-1}. \end{aligned} \tag{25}$$

Step 4. For $i = 0, 1, \dots, m$; $j = 0, 1, \dots, n$, compute $a_{i,j} = \varphi_{i,j}(x_0, \dots, x_i, y_0, \dots, y_j)$.

Step 5. For $i = 0, 1, \dots, m$; $j = 1, \dots, n$ judge if $a_{i,j} \neq 0$; if yes go to Step 6, otherwise exist and show “the algorithm is not valid.”

Step 6. For $i = 0, 1, \dots, m$; $j = n - 2, n - 3, \dots, 0$

$$\begin{aligned} Q_{i,n} &= 1, \quad P_{i,n} = a_{i,n}, \quad Q_{i,n-1} = \|a_{i,n}\|^2, \\ P_{i,n-1} &= a_{i,n-1}Q_{i,n-1} + (y - y_{n-1})\overline{P_{i,n}}, \\ Q_{i,j} &= \|a_{i,j+1}\|^2 Q_{i,j+1} + 2(y - y_{j+1}) \operatorname{tr}(a_{i,j+1} \overline{P_{i,j+2}}) \\ & \quad + (y - y_{j+1})^2 Q_{i,j+2}, \\ P_{i,j} &= a_{i,j}Q_{i,j} + (y - y_j)\overline{P_{i,j+1}}. \end{aligned} \tag{26}$$

Step 7. For $i = 0, 1, \dots, m$; $j = n, n - 1, \dots, 1$, let $B_{i,j}(y) = P_{i,j}(y)/Q_{i,j}(y)$; if $B_{i,j}(y_{j-1}) \neq 0$ then go to Step 8, otherwise exist and show “the algorithm is not valid.”

Step 8. For $i = 0, 1, \dots, m$, let $\widehat{B}_i(y) = B_{i,0}(y) = P_{i,0}(y)/Q_{i,0}(y)$.

Step 9. For $k = 1, \dots, m$ let $\widehat{R}_0(x, y) = \widehat{B}_0(y)$, then compute $\widehat{R}_k(x, y) = \widehat{R}_{k-1}(x, y) + \dots + (x - x_0) \cdots (x - x_{k-1}) \widehat{B}_k(y)$.

3. Some Properties

Lemma 8 (see [5]). Define $U_0(y) = \varphi_{0,0}(x_0, y_0) + (y - y_0)/\varphi_{0,1}(x_0, y_0, y_1) + \dots + (y - y_{n-1})/\varphi_{0,n}(x_0, y_0, \dots, y_n)$; if we use generalized inverse by a tail-to-head rationalization, then a polynomial matrix $E_0(y)$ and a real polynomial $F_0(y)$ exist such that

- (i) $U_0(y) = E_0(y)/F_0(y), F_0(y) \geq 0$,
- (ii) $F_0(y) \mid \|E_0(y)\|^2$, “ \mid ” means the sign of divisibility.

Definition 9. $R(x, y) = N(x, y)/D(x, y)$ is defined to be of type $[l/t]$ if $\deg\{N_{ij}\} \leq l$ for $1 \leq i \leq u, 1 \leq j \leq v, \deg\{N_{ij}\} = l$ for some (i, j) and $\deg\{D\} = t$, where $N(x, y) = (N_{ij}(x, y)) \in C^{u \times v}$.

Lemma 10 (see [5]). Let $U_0(y)$ be defined as in Lemma 8, and $U_0(y) = E_0(y)/F_0(y)$; if n is even, $U_0(y)$ is of type $[n/n]$; if n is odd, $U_0(y)$ is of type $[n/n - 1]$.

Theorem 11. Let $\varphi_{l,k} \in C^{u \times v}, (x_i, y_k) \in \Lambda_{m,n}$, and $x, y \in R$. Define $U_i(y)$ of $R_{mn}(x, y)$ as in (7) by a tail-to-head rationalization using generalized inverse and suppose every intermediate denominator be nonzero in the operation, then a square polynomial matrix $N(x, y)$ and a real polynomial $D(x, y)$ exist such that

- (i) $R_{m,n}(x, y) = N(x, y)/D(x, y)$,
- (ii) $D(x, y) \geq 0$,
- (iii) $D(x, y) \mid \|N(x, y)\|^2$.

Proof. Consider the following algorithm for the construction of $N(x, y), D(x, y)$, and $R_{mn}(x, y)$.

Initialization: let $D_0(x, y) = 1$, and

$$N_0(x, y) = U_0(y) = \varphi_{0,0} + \frac{y - y_0}{\varphi_{0,1}} + \dots + \frac{y - y_{n-1}}{\varphi_{0,n}}. \tag{27}$$

By Lemma 8, a polynomial matrix $E_0(y)$ and a real polynomial $F_0(y)$ exist such that

- (a1) $N_0(x, y) = E_0(y)/F_0(y)$,
- (a2) $F_0(y) \geq 0$,
- (a3) $F_0(y) \mid \|E_0(y)\|^2$.

Recursion: for $j = 1, 2, \dots, m$ let

$$S^{(j)}(x, y) = S^{(j-1)}(x, y) + U_j(y)(x - x_0) \cdots (x - x_{j-1}) \tag{28}$$

with $U_j(y) = E_j(y)/F_j(y), F_j(y) \mid \|E_j(y)\|^2$ having the representation

- (b1) $S^{(j)}(x, y) = N_j(x, y)/D_j(x, y)$,
- (b2) $D_j(x, y) \geq 0$,
- (b3) $D_j(x, y) \mid \|N_j(x, y)\|^2$,

where $D_j(x, y)$ is a real polynomial.

Then we can get

$$\begin{aligned}
 S^{(j+1)}(x, y) &= S^{(j)}(x, y) + U_{j+1}(x - x_0) \cdots (x - x_j) \\
 &= \frac{N_j(x, y)}{D_j(x, y)} + \frac{E_{j+1}(y)}{F_{j+1}(y)} (x - x_0) \cdots (x - x_j) \\
 &= (N_j(x, y) F_{j+1}(y) + D_j(x, y) E_{j+1}(y) \\
 &\quad \times (x - x_0) \cdots (x - x_j)) \\
 &\quad \times (D_j(x, y) F_{j+1}(y))^{-1} \\
 &= \frac{N_{j+1}(x, y)}{D_{j+1}(x, y)},
 \end{aligned} \tag{29}$$

where

$$\begin{aligned}
 N_{j+1}(x, y) &= N_j(x, y) F_{j+1}(y) \\
 &\quad + D_j(x, y) E_{j+1}(y) (x - x_0) \\
 &\quad \cdots (x - x_j), \\
 D_{j+1}(x, y) &= D_j(x, y) F_{j+1}(y).
 \end{aligned} \tag{30}$$

Obviously,

$$\begin{aligned}
 \|N_{j+1}\|^2 &= \|N_j(x, y)\|^2 F_{(j+1)}^2(y) + D_j^2(x, y) \\
 &\quad \times \|E_{(j+1)}(y)\|^2 (x - x_0)^2 \cdots (x - x_j)^2 \\
 &\quad + N_j(x, y) \cdot \bar{E}_{j+1}(y) F_{j+1}(y) D_j(x, y) \\
 &\quad \times (x - x_0) \cdots (x - x_j) \\
 &\quad + \bar{N}_j(x, y) \cdot E_{j+1}(y) F_{j+1}(y) D_j(x, y) \\
 &\quad \times (x - x_0) \cdots (x - x_j)
 \end{aligned} \tag{31}$$

since

$$\begin{aligned}
 D_j(x, y) \mid \|N_j(x, y)\|^2, \quad F_{j+1}(y) \mid \|E_{j+1}(y)\|^2, \\
 D_{j+1}(x, y) \mid \|N_{j+1}(x, y)\|^2.
 \end{aligned} \tag{32}$$

Termination: $S^{(m)}(x, y) = R_{m,n}(x, y)$. □

Theorem 12 (Characterization). *Let $R_{mn}(x, y) = N(x, y)/D(x, y)$ be expressed as*

$$\begin{aligned}
 R = R_{m,n}(x, y) &= U_0(y) + U_1(y)(x - x_0) \\
 &\quad + \cdots + U_m(y)(x - x_0) \cdots (x - x_{m-1}),
 \end{aligned} \tag{33}$$

where $U_i(y) = E_i/F_i$, as in (12) we assume F_i, F_j have no common factor, $i, j = 0, 1, \dots, m, i \neq j$, then

- (i) if n is even, R is of type $[mn + m + n/mn + n]$;
- (ii) if n is odd, R is of type $[mn + n/mn + n - m - 1]$.

Proof. The proof is recursive. Let n be even.

For $m = 0$, $R_{0,n}(x, y) = U_0(y) = E_0/F_0$, by Lemma 10, we find that it is of type $[n/n]$.

For $m = 1$, because $\deg\{E_0\} = \deg\{F_0\} = n$, $\deg\{E_1\} = \deg\{F_1\} = n$

$$\begin{aligned}
 R_{1,n}(x, y) &= \frac{N_1}{D_1} = U_0(y) + U_1(y)(x - x_0) \\
 &= \frac{E_0}{F_0} + \frac{E_1}{F_1}(x - x_0) \\
 &= \frac{E_0 F_1 + E_1 F_0 (x - x_0)}{F_0 F_1}
 \end{aligned} \tag{34}$$

is of type $[2n + 1/2n]$.

For $m = 2$, from the formula above we know that $\deg\{N_1\} = 2n + 1$, $\deg\{D_1\} = 2n$, $\deg\{E_2\} = n$, and $\deg\{F_2\} = n$. one has

$$\begin{aligned}
 R_{2,n}(x, y) &= \frac{N_2}{D_2} = \frac{N_1}{D_1} + \frac{E_2}{F_2}(x - x_0)(x - x_1) \\
 &= \frac{N_1 F_2 + E_2 D_1 (x - x_0)(x - x_1)}{D_1 F_2}.
 \end{aligned} \tag{35}$$

It is easy to find that $R_{2,n}(x, y)$ is of type $[3n + 2/3n]$.

For $m = k$, let $R_{k,n}(x, y) = N_k/D_k$ be of type $[(k + 1)n + k/(k + 1)n]$.

For $m = k + 1$, we consider that

$$\begin{aligned}
 R_{k+1,n}(x, y) &= \frac{N_k}{D_k} + \frac{E_{k+1}}{F_{k+1}}(x - x_0) \cdots (x - x_k) \\
 &= \frac{N_k F_{k+1} + E_{k+1} D_k (x - x_0) \cdots (x - x_k)}{D_k F_k} \\
 &= \frac{N_{k+1}}{D_{k+1}},
 \end{aligned} \tag{36}$$

where

$$\begin{aligned}
 \deg\{N_{k+1}\} &= (k + 1)(n + 1) + n \\
 &= (k + 2)n + (k + 1), \\
 \deg\{D_{k+1}\} &= (k + 1)n + n = (k + 2)n.
 \end{aligned} \tag{37}$$

Therefore, when n is even, R is of type $[mn + m + n/mn + n]$.

Similarly, if n is odd, for $m = 0$, $R_{0,n}(x, y) = U_0(y) = E_0/F_0$, by Lemma 10, we find that it is of type $[n/n - 1]$.

For $m = 1$, because $\deg\{E_0\} = \deg\{E_1\} = n$, $\deg\{F_0\} = \deg\{F_1\} = n - 1$

$$\begin{aligned}
 R_{1,n}(x, y) &= \frac{N_1}{D_1} = U_0(y) + U_1(y)(x - x_0) \\
 &= \frac{E_0}{F_0} + \frac{E_1}{F_1}(x - x_0) \\
 &= \frac{E_0 F_1 + E_1 F_0 (x - x_0)}{F_0 F_1}
 \end{aligned} \tag{38}$$

is of type $[2n/2n - 2]$.

For $m = 2$, from the formula above we know that $\deg\{N_1\} = 2n$, $\deg\{D_1\} = 2n - 2$, $\deg\{E_2\} = n$, and $\deg\{F_2\} = n - 1$. one has

$$R_{2,n}(x, y) = \frac{N_2}{D_2} = \frac{N_1}{D_1} + \frac{E_2}{F_2}(x - x_0)(x - x_1) \tag{39}$$

$$= \frac{N_1 F_2 + E_2 D_1 (x - x_0)(x - x_1)}{D_1 F_2}.$$

We get that $R_{2,n}(x, y)$ is of type $[3n/3n - 3]$.

For $m = k$, let

$$R_{k,n}(x, y) = \frac{N_k}{D_k} \tag{40}$$

be of type $[(k + 1)n/(k + 1)(n - 1)]$.

For $m = k + 1$, we consider that

$$R_{k+1,n}(x, y) = \frac{N_k}{D_k} + \frac{E_{k+1}}{F_{k+1}}(x - x_0) \cdots (x - x_k) \tag{41}$$

$$= \frac{N_k F_{k+1} + E_{k+1} D_k (x - x_0) \cdots (x - x_k)}{D_k F_{k+1}}$$

$$= \frac{N_{k+1}}{D_{k+1}},$$

where

$$\deg\{N_{k+1}\} = (k + 1)(n - 1) + k + 1 = (k + 2)n,$$

$$\deg\{D_{k+1}\} = (k + 1)(n - 1) + (n - 1) = (k + 2)(n - 1). \tag{42}$$

Therefore, when n is odd, R is of type $[(m + 1)n/(m + 1)(n - 1)]$. \square

Definition 13. A matrix-valued Newton-Thiele type rational fraction $R_{m,n}(x, y) = N(x, y)/D(x, y)$ is defined to be a bivariate generalized inverse and rational interpolant (BGIRI_{NT}) on the rectangular grid $\Lambda_{m,n}$ if

- (i) $R_{mn}(x_l, y_k) = Y_{lk}, (x_l, y_k) \in \Lambda_{m,n}$,
- (ii) $D(x_l, y_k) \neq 0, (x_l, y_k) \in \Lambda_{m,n}$,
- (iii) (a) if n is even,

$$\deg\{N(x, y)\} = mn + m + n, \quad \deg\{D(x, y)\} = mn + n, \tag{43}$$

- (b) if n is odd,

$$\deg\{N(x, y)\} = mn + n, \quad \deg\{D(x, y)\} = mn + n - m - 1, \tag{44}$$

- (iv) $D(x, y) \mid \|N(x, y)\|^2$,

- (v) $D(x, y)$ is real, and $D(x, y) \geq 0$.

Now let us turn to the error estimate of the BGIRI_{NT}. Firstly, we give Lemma 14.

Lemma 14. Let

$$R_{m,n}^*(x, y) = U_0^*(y) + U_1^*(y)(x - x_0) + \cdots + U_m^*(y)(x - x_0)(x - x_1) \cdots (x - x_{m-1}), \tag{45}$$

where for $l = 0, 1, \dots, m$

$$U_l^*(y) = \varphi_{l,0}(x_0, \dots, x_l, y_0) + \frac{y - y_0}{\varphi_{l,1}(x_0, \dots, x_l, y_0, y_1)} + \cdots + \frac{y - y_{n-1}}{|\varphi_{l,n}(x_0, \dots, x_l, y_0, \dots, y_n)|} + \frac{y - y_n}{|\varphi_{l,n+1}(x_0, \dots, x_l, y_0, \dots, y_n, y)|}, \tag{46}$$

then $R_{m,n}^*(x, y)$ satisfies

$$R_{m,n}^*(x_i, y) = f(x_i, y), \quad i = 0, 1, \dots, m, \tag{47}$$

$$R_{m,n}^*(x, y_j) = R_{m,n}(x, y_j), \quad j = 0, 1, \dots, n.$$

Remark 15. We delete the proof of the above lemma since it can be easily generalized from [3].

Now the error estimate Theorem 16 will be put which is also motivated by [3]. We let

$$R_{m,n}(x, y) = \frac{N(x, y)}{D(x, y)}, \quad R_{m,n}^*(x, y) = \frac{N^*(x, y)}{D^*(x, y)}. \tag{48}$$

Theorem 16. Let a matrix $f(x, y)$ be $\max(m + 1, n + 1)$ times continuously differentiable on a set $\bar{D} = \{a \leq x \leq b, c \leq y \leq d\}$ containing the points $\Lambda_{m,n} = \{(x_l, y_k) : l = 0, 1, \dots, m, k = 0, 1, \dots, n, (x_l, y_k) \in \mathbb{R}^2\}$, then, for any point $(x, y) \in \bar{D}$, there exists a point $(\xi, \eta) \in \bar{D}$ such that

$$f(x, y) - R_{m,n}(x, y) = \frac{1}{D(x, y)D^*(x, y)} \times \left\{ \frac{\omega_{m+1}(x)}{(m + 1)!} \frac{\partial^{m+1}}{\partial x^{m+1}} E_1(\xi, y) + \frac{\omega_{n+1}(y)}{(n + 1)!} \frac{\partial^{n+1}}{\partial y^{n+1}} E_2(x, \eta) \right\}, \tag{49}$$

where

$$\omega_{m+1}(x) = (x - x_0)(x - x_1) \cdots (x - x_m),$$

$$\omega_{n+1}(y) = (y - y_0)(y - y_1) \cdots (y - y_n),$$

$$E_1(x, y) = D(x, y)D^*(x, y) [f(x, y) - R_{m,n}^*(x, y)],$$

$$E_2(x, y) = D(x, y)D^*(x, y) [R_{m,n}^*(x, y) - R_{m,n}(x, y)]. \tag{50}$$

Proof. Let

$$\begin{aligned} E_1(x, y) &= D(x, y) D^*(x, y) [f(x, y) - R_{m,n}^*(x, y)], \\ E_2(x, y) &= D(x, y) D^*(x, y) [R_{m,n}^*(x, y) - R_{m,n}(x, y)], \\ E(x, y) &= E_1(x, y) + E_2(x, y). \end{aligned} \tag{51}$$

From Lemma 14, we know

$$E_1(x_i, y) = 0, \quad i = 0, 1, \dots, m, \tag{52}$$

which results in

$$E_1(x, y) = \frac{\omega_{m+1}(x)}{(m+1)!} \frac{\partial^{m+1}}{\partial x^{m+1}} E_1(x, y) \Big|_{x=\xi}, \tag{53}$$

where $\omega_{m+1}(x) = (x - x_0)(x - x_1) \cdots (x - x_m)$, and ξ is a number contained in the interval (a, b) which may depend on y . Similarly, from

$$E_2(x, y_j) = 0, \quad j = 0, 1, \dots, n, \tag{54}$$

we can get that

$$E_2(x, y) = \frac{\omega_{n+1}(y)}{(n+1)!} \frac{\partial^{n+1}}{\partial y^{n+1}} E_2(x, y) \Big|_{y=\eta}, \tag{55}$$

where $\omega_{n+1}(y) = (y - y_0)(y - y_1) \cdots (y - y_n)$, and η is a number contained in the interval (c, d) which may depend on x .

Therefore,

$$\begin{aligned} E(x, y) &= E_1(x, y) + E_2(x, y) \\ &= \frac{\omega_{m+1}(x)}{(m+1)!} \frac{\partial^{m+1}}{\partial x^{m+1}} E_1(x, y) \Big|_{x=\xi} \\ &\quad + \frac{\omega_{n+1}(y)}{(n+1)!} \frac{\partial^{n+1}}{\partial y^{n+1}} E_2(x, y) \Big|_{y=\eta}. \end{aligned} \tag{56}$$

The proof is thus completed. □

4. Numerical Examples

Example 17. Let (x_i, y_j) and $Y_{i,j}, i, j = 0, 1, 2$ be given in Table 1:

$$R_{2,2}(x, y) = U_0(y) + U_1(y)x + U_2(y)x(x-1). \tag{57}$$

Using Algorithm 7, we can get

$$\begin{aligned} U_0(y) &= \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + \frac{y}{\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}} + \frac{y-1}{\begin{vmatrix} 0 & 0 \\ -2/5 & -1/5 \end{vmatrix}}, \\ U_1(y) &= \begin{pmatrix} 0 & -1 \\ 0 & 0 \end{pmatrix} + \frac{y}{\begin{pmatrix} 0 & 1/2 \\ -1/2 & 0 \end{pmatrix}} + \frac{y-1}{\begin{vmatrix} 0 & 1 \\ 1 & 0 \end{vmatrix}}, \\ U_2(y) &= \begin{pmatrix} 0 & 1 \\ 1/2 & 0 \end{pmatrix} + \frac{y}{\begin{pmatrix} 0 & -3/5 \\ 1/5 & 0 \end{pmatrix}} + \frac{y-1}{\begin{vmatrix} 0 & -15/14 \\ -5/14 & -5/7 \end{vmatrix}}. \end{aligned} \tag{58}$$

Based on generalized inverse (4), we obtain

$$\begin{aligned} R_{2,2}(x, y) &= \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \times (70y^6 - 368y^5 + 892y^4 \\ &\quad - 1240y^3 + 1040y^2 - 512y + 128)^{-1} \\ &= \frac{N(x, y)}{D(x, y)}, \end{aligned} \tag{59}$$

where

$$\begin{aligned} a_{11} &= 70y^6 - 368y^5 + 892y^4 - 1240y^3 \\ &\quad + 1040y^2 - 512y + 128, \\ a_{12} &= 178y^5x - 858y^4x + 1688y^3x \\ &\quad - 1776xy^2 + 1024xy - 256x - 38y^5x^2 \\ &\quad + 262y^4x^2 - 640y^3x^2 - 512x^2y \\ &\quad - 5y^6x^2 + 5y^6x + 800x^2y^2 + 128x^2, \\ a_{21} &= 108y^5 - 200y^4 + 200y^3 - 112y^2 + 32y \\ &\quad - 28y^6 + 60y^6x - 344y^5x + 736y^4x - 784y^3x \\ &\quad + 408xy^2 - 32xy - 64x - 24y^5x^2 + 16y^4x^2 \\ &\quad - 96x^2y + 10y^6x^2 + 40x^2y^2 + 64x^2, \\ a_{22} &= 68y^5 - 140y^4 + 160y^3 - 96y^2 \\ &\quad + 32y - 14y^6 - 50y^6x^2 + 270y^5x^2 - 640y^4x^2 \\ &\quad + 820y^3x^2 - 560x^2y^2 + 160x^2y + 50y^6x - 270y^5x \\ &\quad + 640y^4x - 820y^3x + 560xy^2 - 160xy. \end{aligned} \tag{60}$$

From Definition 13, we know that $R_{2,2}(x, y)$ is a BGIRI_{NT} since $R_{2,2}(x_i, y_j) = Y_{i,j}$ and $\deg N(x, y) = 8, \deg D(x, y) = 6$, and $D(x, y) \mid \|N(x, y)\|^2$.

In paper [9], a bivariate Thieletype matrix-valued rational interpolant $R_T(x, y) = G_{0,0}(y) + (x - x_0)/G_{1,0}(y) + \cdots + (x - x_{m-1})/G_{n,0}(y)$ (page 73), where for $l = 0, 1, \dots, n$,

$$\begin{aligned} G_{l,0}(y) &= B_{l,0}(x_0, \dots, x_l, y_0) \\ &\quad + \frac{y - y_0}{B_{l,1}(x_0, \dots, x_l, y_0, y_1)} \\ &\quad + \cdots + \frac{y - y_{m-1}}{|B_{l,m}(x_0, \dots, x_l, y_0, \dots, y_m)|}. \end{aligned} \tag{61}$$

Here $B_{l,k}(x_0, \dots, x_l, y_0, \dots, y_k), l = 0, 1, \dots, n, k = 0, 1, \dots, m$ is defined different from that of this paper (see [9]). We

TABLE 1

y_j/x_i	$x_0 = 0$	$x_1 = 1$	$x_2 = 2$
$y_0 = 0$	$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$	$\begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix}$	$\begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix}$
$y_1 = 1$	$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$	$\begin{pmatrix} 1 & 0 \\ -1 & 1 \end{pmatrix}$	$\begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix}$
$y_2 = 2$	$\begin{pmatrix} 1 & 0 \\ -1 & 0 \end{pmatrix}$	$\begin{pmatrix} 1 & 1 \\ -1 & 0 \end{pmatrix}$	$\begin{pmatrix} 1 & 1 \\ 0 & -1 \end{pmatrix}$

TABLE 2

$Y(1.3, 1.3)$	$R_{NT}(1.3, 1.3)$	$\ R_{NT} - Y\ _F$
$\begin{pmatrix} -0.856888 & 0.5155013 & 3.669296 \\ 1.30000 & 3.66929 & 2.600 \end{pmatrix}$	$\begin{pmatrix} -0.856888 & 0.5155010 & 3.669297 \\ 1.30000 & 3.66929 & 2.600 \end{pmatrix}$	$1.09e - 006$
	$R_T(1.3, 1.3)$	$\ R_T - Y\ _F$
	$\begin{pmatrix} -0.98999 & 0.14111 & 5.47394 \\ 1.29999 & 5.47394 & 2.999 \end{pmatrix}$	2.613697
$Y(1.5, 1.5)$	$R_{NT}(1.5, 1.5)$	$\ R_{NT} - Y\ _F$
$\begin{pmatrix} -0.989999 & 0.1411200 & 4.481689 \\ 1.50000 & 4.481689 & 3.00 \end{pmatrix}$	$\begin{pmatrix} -0.989992 & 0.1411203 & 4.481688 \\ 1.50000 & 4.481688 & 3.00 \end{pmatrix}$	$9.15e - 007$
	$R_T(1.5, 1.5)$	$\ R_T - Y\ _F$
	$\begin{pmatrix} -0.99829372 & -0.05837 & 5.47394 \\ 1.49999 & 5.47394 & 3.1999 \end{pmatrix}$	1.43144

now give another numerical example to compare the two algorithms, which shows that the method of this paper is better than the one of [9].

Example 18. Let

$$Y(x, y) = \begin{pmatrix} \cos(x + y) & \sin(x + y) & e^y \\ x & e^y & x + y \end{pmatrix} \quad (62)$$

and we suppose $\{x_0, x_1, x_2, x_3, x_4, x_5\} = \{y_0, y_1, y_2, y_3, y_4, y_5\} = \{1.0, 1.2, 1.4, 1.6, 1.8, 2.0\}$. The numerical results $R_{NT}(x, y)$ in Step 9 of Algorithm 7 and $R_T(x, y)$ in [9] are given in Table 2.

Remark 19. In Table 2, from the F-norm of $R_{NT} - Y$ and $R_T - Y$, we can see that the error using Newton-Thiele type formula is much less than the one using Thieletype formula.

5. Conclusion

In this paper, the bivariate generalized inverse Newton-Thiele type matrix interpolation to approximate a matrix function $f(x, y)$ is given, and we also give a recursive algorithm accompanied with some other important conclusions such as divisibility, characterization and some numerical examples. From the second example, it is easy to find that the approximant using Newton-Thiele type formula is much better than the one using Thieletype formula.

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