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The refinement-schemes-based unified algorithms for certain *n*th order linear and nonlinear differential equations with a set of constraints

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Abstract

We first present a generalized class of binary interpolating refinement schemes and their properties. Then the refinement-schemes-based unified algorithms for the solution of certain *n*th order linear and nonlinear differential equations with a set of constraints are presented. Moreover, several algorithms based on the refinement schemes for solving differential equations are the special cases of our algorithms.

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1 Introduction

The refinement schemes, also known as subdivision schemes, are efficient tools for the modeling of curves. These schemes are classified into two main categories, interpolating and approximating. These categories are further classified into *n* subcategories: binary, ternary, ..., *n*-ary. In this paper, we focus on binary interpolating refinement schemes. The domain of these schemes is a polygon while the range is a refined polygon. These schemes have two main rules, namely refinement and topological rules. There are two refinement rules: one rule caries on the points of a coarse polygon while the other rule introduces the new points corresponding to each edge of the polygon. These rules are called even and odd rules, respectively. The even rule just caries on the old points. The odd rule is an affine combination of the points of the coarse polygon. Furthermore, the topological rule is just the connection of adjacent new and old points with straight lines. The topological and refinement rules give us a new polygon. The repeated application of the refinement and topological rules is a smooth shape. Graphically, this procedure is depicted in Fig. 1. Mathematically, if $Q^k = \{Q_i^k\}_{i \in \mathbb{Z}}$ is a polygon at the *k*th level then the refined polygon $Q^{k+1} =$

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 $\{Q_i^{k+1}\}_{i\in\mathbb{Z}}$ can be obtained by applying the topological and refinement rules [1] as follows:

(Even rule)
$$Q_{2i}^{k+1} = Q_i^k$$
,
(Odd rule) $Q_{2i+1}^{k+1} = \sum_{i=0}^{n_1} \chi_{n_1,i} (Q_{i-i}^k + Q_{i+i+1}^k)$, (1)

where the coefficients appearing in the affine combination of points are

$$\chi_{n_{1},j} = \frac{((2n_{1}+1)!!)^{2}}{2(4^{n_{1}})(2n_{1}+1)!} \frac{(-1)^{j}}{(2j+1)} \binom{2n_{1}+1}{n_{1}-j}, \quad j = 0, 1, 2, \dots, n_{1},$$
(2)

while $\binom{2n_1+1}{n_1-j}$ denotes the binomial coefficient. Here n_1 is the complexity of the scheme. If $n_1 = 0, 1, 2, 3, ...$, then the complexity of the scheme will be 2, 4, 6, 8, ..., respectively. In other words, for $n_1 = 0, 1, 2, 3$, we get 2-, 4-, 6-, and 8-point schemes, respectively.

The refinement procedure has attracted attention due to a large variety of applications in curve modeling and algorithms for the solution of differential equations with a set of constraints. Mathematically, these equations are called the boundary value problems (BVPs). Higher-order linear and nonlinear differential equations have been reported in mathematical physics and structural engineering. Different techniques have been introduced to solve such problems. Here is a list of works that have caught the attention of the scientific community, pointing to the diversity of the applications of refinement schemes in the area of differential equations.

In 1996, initially, Qu and Agarwal [2] presented a refinement-scheme-based algorithm for the second-order linear differential equations (DEs). A year later, Qu and Agarwal [3] offered an algorithm for the second-order nonlinear DEs. Then after the long silence, Mustafa and Ejaz [4] introduced the refinement-based algorithm for the third-order linear DEs in 2014. In 2015, Ejaz et al. [5] introduced an algorithm for the fourth-order linear DEs. Ejaz and Mustafa [6] offered an algorithm for the third-order nonlinear DEs in 2016. In the next year, Mustafa et al. [7] introduced an algorithm for the fourth-order nonlinear DEs.

In this paper, we present generalized algorithms based on generalized refinement schemes for the *n*th order linear and nonlinear DEs. We prove that all the above algorithms are special cases of our generalized algorithms. We consider the following two-point *n*th order linear and nonlinear DEs with a set of constraints

(Linear DE)
$$u^n(t) + q(t)u(t) = f(t), \quad a \le t \le b,$$
 (3)

and

(Nonlinear DE)
$$u^n(t) = f(t, u(t), u'(t)),$$
 (4)

where the set of constraints is defined as follows: If n is even, a set of constraints is

$$u^{l}(a) = \alpha_{l}, \qquad u^{m}(b) = \beta_{m}, \quad l, m = 0, 1, 2, \dots, n-2;$$
(5)

If *n* is odd, a set of constraints is

$$u^{l}(a) = \alpha_{l}, \qquad u^{m-1}(b) = \beta_{m}, \quad l, m = 0, 1, 2, \dots, n-2,$$

or
$$u^{l-1}(a) = \alpha_{l}, \qquad u^{m}(b) = \beta_{m}, \quad l, m = 0, 1, 2, \dots, n-2,$$

(6)

where α_l and β_m are scalars. We assume that the problems are well-posed throughout the paper.

The rest of the work is structured as follows. In Sect. 2, we discuss the properties of generalized binary interpolating refinement schemes. We also present generalized formulae for the *n*th derivatives of the refinement schemes in this section. The generalized algorithms for the *n*th order linear and nonlinear DEs are presented in Sect. 3. In Sect. 4, we present the generalized form of imposed constraints and the approximation of derivative involved in the constraints. In Sect. 5, we discuss the stable linear and nonlinear system of equations. We also discuss the existence of the solutions of these systems in this section. In Sect. 6, we show that the refinement-based existing algorithms are special cases of our generalized algorithms. Section 7 presents the conclusion.

2 Properties of the refinement scheme

The *n*th order continuous (i.e., C^n continuous) refinement scheme is suitable to develop an algorithm for the solution of the *n*th order DEs. For example, if we want to find solutions of the eighth order DEs with a set of constraints then we have to choose a C^8 -continuous refinement scheme from (1).

Here we briefly summarize the continuity and other properties of a refinement scheme. If $\{Q_i = (i, \delta_0)^T\}$ is the initial data then repeated application of the scheme produces the limit curve named $\rho(t)$, also known as a basis function, where

$$\rho(i) = \begin{cases}
1, & i = 0, \\
0, & i \neq 0.
\end{cases}$$
(7)

and

$$\rho(t) = \rho_{n_1}(t) = \rho(2t) + \sum_{j=-n_1}^{n_1} \chi_{n_1,|j|} \rho(2t - 2j + 1), \quad t \in \mathbb{R}.$$
(8)

The scheme (1) has the following properties:

- It produces C^n continuous curves, where $n = n_1$ for $n_1 \le 5$, and for a large value of n_1 , $n = 0.415n_1$ by [8]. It means that $\rho(t)$ is n times continuously differentiable.
- Its degree of generation and reproduction is $2n_1 + 1$.
- The approximation order of the scheme is $2n_1 + 2$.
- The support of $\rho(t)$ is finite; explicitly, its support is $(-2n_1 1, 2n_1 + 1)$.

	$\begin{pmatrix} 0 \end{pmatrix}$	0	0	•••	0	1	0	0	0	•••
	χ_{n_1,n_1}	χ_{n_1,n_1-1}	χ_{n_1,n_1-2}	•••	$\chi_{n_1,-1}$	$\chi_{n_1,0}$	$\chi_{n_1,0}$	$\chi_{n_1,-1}$	$\chi_{n_1,-2}$	•••
<i>S</i> =	0	0	0	•••	0	0	1	0	0	•••
	0	χ_{n_1,n_1}	χ_{n_1,n_1-1}	•••	$\chi_{n_1,-2}$	$\chi_{n_1,-1}$	$\chi_{n_1,0}$	$\chi_{n_1,0}$	$\chi_{n_1,-1}$	•••
	0	0	0	•••	0	0	0	1	0	•••
	0	0	χ_{n_1,n_1}	•••	$\chi_{n_1,-3}$	$\chi_{n_1,-2}$	$\chi_{n_1,-1}$	$\chi_{n_1,0}$	$\chi_{n_1,0}$	•••
	0	0	0	•••	0	0	0	1	0	•••
	:	÷	:	÷	÷	:	:	÷	÷	÷
	0	0	0		0	0	0	0	0	
	0	0	0	•••	0	0	0	0	0	•••
	0	0	0	0		0	())		
	χ_{n_1,n_1-2}	χ_{n_1,n_1-1}	χ_{n_1,n_1}	0		0	(0		
	0	0	0	0		0	(0		
	χ_{n_1,n_1-3}	χ_{n_1,n_1-2}	χ_{n_1,n_1-1}	χ_{n_1}	, <i>n</i> 1 ·	0	(0		
		0	0	0		0	(D		(0)
	χ_{n_1,n_1-4}	χ_{n_1,n_1-3}	χ_{n_1,n_1-2}	<i>χn</i> 1, <i>n</i>	i_1-1 .	0	(0 ·		(9)
	•••	0	0	0		0	(D		
	:	÷	:	÷		: :		:		
	0	0	0	0		·· χ _{n1,n}	₁₋₁ χ _n	1, <i>n</i> 1		
	0	0	0	0		0	(o /		

• The iteration matrix, also known as a local refinement matrix, is defined in (9).

The order of the refinement matrix is $4n_1 + 1$.

The matrix S has the following k eigenvalues λ_k and their corresponding right-eigenvectors R_k:

$$\lambda_{k} = 2^{-k}, \quad 0 \le k \le 2n_{1} + 1,$$

$$R_{k} = \left((-2n_{1})^{k}, (-2n_{1} + 1)^{k}, \dots, (-2)^{k}, (-1)^{k}, 0, 1, 2^{k}, \dots, (2n_{1} - 1)^{k}, (2n_{1})^{k} \right)_{(4n_{1} + 1) \times 1}^{T}.$$
(10)

• The left-eigenvectors L_k corresponding to the eigenvalues λ_k can be found by using the transpose of local refinement matrix *S*. These eigenvectors also satisfy the relation $R_i^T L_j = \delta_i \delta_j, \forall i, j: i, j = 1, 2, ..., n_1$. That is,

$$R_i^T L_j = 1, \quad \text{if } i = j,$$
$$R_i^T L_j = 0, \quad \text{if } i \neq j.$$

Proposition 1 The function $\rho(t)$ is n-times continuously differentiable on the interval $(-2n_1 - 1, 2n_1 + 1)$ and its nth derivatives are given by

$$\rho^{n}(t) = 2^{n} (\operatorname{sgn}(t))^{n} E_{|t|}^{T} L_{n}, \quad -2n_{1} \leq t \leq 2n_{1},$$

where

$$\operatorname{sgn}(t) = \begin{cases} -1 & \text{if } t < 0, \\ 0 & \text{if } t = 0, \\ 1 & \text{if } t > 0, \end{cases}$$

 $E_t = (a_{(2n)t}, a_{(2n-1)t}, \dots, a_{1t}, a_{0t}, a_{-1t}, a_{-2t}, \dots, a_{(-2n+2)t}, a_{(-2n+1)t}, a_{(-2n)t}),$

for $0 \le t \le 4n_1 + 1$ *and*

$$a_{pt} = \begin{cases} 1 & if p = t, \\ 0 & if p \neq t. \end{cases}$$

3 Generalized algorithms for the *n*th order linear and nonlinear DEs

We structure the refinement-schemes-based algorithms for the two-point nth order linear and nonlinear DEs with the set of constraints (3) and (4).

3.1 Generalized algorithm for the *n*th order linear DEs

In this section, we construct the refinement-scheme-based algorithm for the nth order linear DEs. Let the solution of (3) be

$$G_{L}(t) = \sum_{i=-2n_{1}}^{N+2n_{1}} g_{i}^{L} \rho\left(\frac{t-t_{i}}{h}\right),$$
(11)

where $a \le t \le b$, $N \ge 2n_1$, $t_i = ih$, $g_i^L = G_L(t_i)$, and $h = \frac{b-a}{N}$. This implies by (3) that

$$G_L^n(t_j) + q(t_j)G_L(t_j) = f(t_j), \quad j = 0, 1, \dots, N.$$
 (12)

If we have a set of constraints as in (5) then we get

$$G_L^l(a) = \alpha_l, \qquad G_L^m(b) = \beta_m, \quad l, m = 0, 1, 2, \dots, n-2.$$
 (13)

For a set of constraints as in (6), it takes the form

$$G_{L}^{l}(a) = \alpha_{l}, \qquad G_{L}^{m-1}(b) = \beta_{m}, \quad l, m = 0, 1, 2, \dots, n-2,$$

or
$$G_{L}^{l-1}(a) = \alpha_{l}, \qquad G_{L}^{m}(b) = \beta_{m}, \quad l, m = 0, 1, 2, \dots, n-2.$$
 (14)

From (11), we have

$$\begin{split} G_L'(t_j) &= \frac{1}{h} \sum_{i=-2n_1}^{N+2n_1} g_i^L \rho' \left(\frac{t_j - t_i}{h} \right), \\ G_L''(t_j) &= \frac{1}{h^2} \sum_{i=-2n_1}^{N+2n_1} g_i^L \rho'' \left(\frac{t_j - t_i}{h} \right), \end{split}$$

÷

$$G_{L}^{n}(t_{j}) = \frac{1}{h^{n}} \sum_{i=-2n_{1}}^{N+2n_{1}} g_{i}^{L} \rho^{n} \left(\frac{t_{j}-t_{i}}{h}\right).$$
(15)

Now using (15) in (12), we get

$$\sum_{i=-2n_1}^{N+2n_1} g_i^L \rho^n \left(\frac{t_j - t_i}{h} \right) + h^n q(t_j) \sum_{i=-2n_1}^{N+2n_1} g_i^L \rho \left(\frac{t_j - t_i}{h} \right) = h^n f(t_j), \quad j = 0, 1, \dots, N.$$

This implies

$$\sum_{i=-2n_1}^{N+2n_1} g_i^L \rho^n(j-i) + h^n q(t_j) \sum_{i=-2n_1}^{N+2n_1} g_i^L \rho(j-i) = h^n f(t_j), \quad j = 0, 1, \dots, N.$$

Let $\rho_i^n = \rho^n(i)$, then

$$\sum_{i=-2n_1}^{N+2n_1} g_i^L \rho_{j-i}^n + h^n q(t_j) \sum_{i=-2n_1}^{N+2n_1} g_i^L \rho_{j-i} = h^n f(t_j), \quad j = 0, 1, \dots, N.$$

This can be simplified as

$$\sum_{i=-2n_1}^{N+2n_1} g_i^L \left[\rho_{j-i}^n + h^n q(t_j) \rho_{j-i} \right] = h^n f(t_j), \quad j = 0, 1, \dots, N.$$
(16)

This system has N + 1 equations. Its matrix is as follows:

$$A_L G_L = d_L, \tag{17}$$

where the banded matrix $(A_L)_{(N+1)\times(N+4n_1+1)}$, column vectors $(G_L)_{(N+4n_1+1)}$ and $(d_L)_{(N+1)}$ are defined in (18), (19), and (20), respectively,

and $\chi_j = \rho_j^n + q_j h^n$,

$$G_L = \left(g_{-2n_1}^L, g_{-2n_1+1}^L, \dots, g_{2n_1+N-1}^L, g_{2n_1+N}^L\right)^T,$$
(19)

and

$$d_L = (h^n f_0, h^n f_1, \dots, h^n f_N)^T.$$
(20)

The system (17) is unstable. To find the solution of (17), we need $4n_1$ more equations. As the *n* constraints are given in (3), so we only need to construct $4n_1 - n$ constraints. The remaining $4n_1 - n$ constraints can be found by some extrapolation method.

For this, we have the following two cases:

- If $4n_1 n$ is even, then we construct $\frac{4n_1-n}{2}$ constraints at the left end of the DE and $\frac{4n_1-n}{2}$ constraints on the right end.
- If $4n_1 n$ is odd then we construct $\frac{4n_1 n + 1}{2}$ constraints at the left end and $\frac{4n_1 n 1}{2}$ constraints at the right end of the DE.

The treatment of these constraints is given in the next section.

3.2 Generalized algorithm for the *n*th order nonlinear DEs

In this subsection, we construct the refinement-scheme-based algorithm for the nth order nonlinear DEs. Let the solution of (4) be

$$G_{NL}(t) = \sum_{i=-2n_1}^{N+2n_1} g_i^{NL} \rho\left(\frac{t-t_i}{h}\right),$$
(21)

where $a \le t \le b$, $N \ge 2n_1$, $t_i = ih$, $g_i^{NL} = G_{NL}(t_i)$, and $h = \frac{b-a}{h}$. This implies by (4) that

$$G_{NL}^{n}(t_{j}) = f\left(t_{j}, G_{NL}(t_{j}), G_{NL}'(t_{j}), \dots, G_{NL}^{n-1}(t_{j})\right), \quad j = 0, 1, \dots, N,$$
(22)

with the set of constraints given in (13) or (14). From (21), we have

$$G'_{NL}(t_{j}) = \frac{1}{h} \sum_{i=-2n_{1}}^{N+2n_{1}} g_{i}^{NL} \rho' \left(\frac{t_{j}-t_{i}}{h}\right),$$

$$G''_{NL}(t_{j}) = \frac{1}{h^{2}} \sum_{i=-2n_{1}}^{N+2n_{1}} g_{i}^{NL} \rho'' \left(\frac{t_{j}-t_{i}}{h}\right),$$

$$\vdots$$

$$G''_{NL}(t_{j}) = \frac{1}{h^{n}} \sum_{i=-2n_{1}}^{N+2n_{1}} g_{i}^{NL} \rho'' \left(\frac{t_{j}-t_{i}}{h}\right).$$
(23)

Now using (23) in (22), we get

$$\sum_{i=-2n_1}^{N+2n_1} g_i^{NL} \rho^n \left(\frac{t_j - t_i}{h} \right) = h^n f(t_j, G_{NL}(t_j), G'_{NL}(t_j), \dots, G_{NL}^{n-1}(t_j)), \quad j = 0, 1, \dots, N.$$

This implies

$$\sum_{i=-2n_1}^{N+2n_1} g_i^{NL} \rho^n(j-i) = h^n f(t_j, G_{NL}(t_j), G'_{NL}(t_j), \dots, G_{NL}^{n-1}(t_j)), \quad j = 0, 1, \dots, N.$$

This leads to

$$\sum_{i=-2n_1}^{N+2n_1} g_i \rho_{j-i}^n = h^n f(t_j, G_{NL,j}, G'_{NL,j}, \dots, G_{NL,j}^{n-1}), \quad j = 0, 1, \dots, N,$$
(24)

where $G_{NL,j}^{l} = G_{NL}^{l}(t_{j})$, for l = 0, 1, ..., n - 1. This system also has N + 1 equations. Its matrix form is

$$A_{NL}G_{NL} = d_{NL}, \tag{25}$$

where A_{NL} is the banded matrix of order $(N + 1) \times (N + 4n_1 + 1)$, G_{NL} and d_{NL} have orders $(N + 4n_1 + 1) \times 1$ and $(N + 1) \times 1$, respectively. These matrices are defined below.

This can be simplified as

$$A_{NL} = (-1)^n \left(\rho_{rc}^n (c - r - 2n_1 - 1) \right),$$

where r = 1, 2, ..., N and $c = 1, 2, ..., N + 4n_1 + 1$. The column matrices are defined as

$$G_{NL} = (g_{-2n_1}, g_{-2n_1+1}, \dots, g_{2n_1+N-1}, g_{2n_1+N})^T$$
(27)

and

$$d_{NL} = h^n (f_0, f_1, \dots, f_{N-1}, f_N)^T,$$
(28)

where $f_j = f(t_j, g_{NL,j}, g'_{NL,j}, \dots, g_{NL,j}^{n-1})$ and $g_{NL,j}^l = g_{NL}^l(t_j)$, for $j = 0, 1, \dots, n$ and $l = 0, 1, \dots, n-1$. To find the solution of (25), we need $4n_1$ more equations to solve the system (25). As the n constraints are given in equation (4), so we only need to construct $4n_1 - n$ constraints. The remaining $4n_1 - n$ constraints can be found by some extrapolation method, the detail is given below.

- If $4n_1 n$ is even, then we construct $\frac{4n_1-n}{2}$ constraints at the left end of the DE and $\frac{4n_1-n}{2}$ constraints on the right end.
- If $4n_1 n$ is odd then we construct $\frac{4n_1 n + 1}{2}$ constraints at the left end and $\frac{4n_1 n 1}{2}$ constraints at the right end of the DE.

4 Approximation of the given and imposed constraints

The given derivative constraints and the imposed constraints for the unstable systems (17) and (25) are approximated in this section.

4.1 Approximation of the given derivative constraints

If F(t) is a function then for h > 0 and integer p > 0, the *l*th derivative of G(t) can be approximated by the finite difference method as

$$G^{l}(t) = \frac{l!}{h^{l}} \sum_{i=i_{\min}}^{i_{\max}} c_{i}F(t+ih) + O(h^{p}).$$
⁽²⁹⁾

The necessary condition for (29) to be satisfied is

$$\sum_{i=i_{\min}}^{i_{\max}} i^n c_i = \begin{cases} 0, & \text{for } 0 \le n \le l+p-1 \text{ and } n \ne l, \\ 1, & \text{for } n=l. \end{cases}$$
(30)

For $i_{\min} = 0$ and $i_{\max} = l + p - 1$, the forward difference approximation can be expected. The convolution mask is the vector $C = (c_{i_{\min}}, \dots, c_{i_{\max}})$. If we solve the system (30) then we get the convolution matrix C.

4.2 Approximation of the imposed constraints

We impose the set of constraints on the left as well as on the right side of the given constraints of the linear and nonlinear DE. These constraints are constructed as follows.

4.2.1 The left end imposed constraints

If $S_1(t)$ is the polynomial which interpolates the data $(t_i, g_i), 0 \le i \le 2n_1 - 1$, then the values $g_{-(2n_1-1)}, g_{-(2n_1)}, \dots, g_{-2}, g_{-1}$ imposed on the left side can be found. That is,

$$g_{-i} = S_1(-t_i), \quad i = \begin{cases} 1, 2, \dots, \frac{4n_1 - n}{2}, & \text{if } 4n_1 - n = \text{even}, \\ 1, 2, \dots, \frac{4n_1 - n + 1}{2}, & \text{if } 4n_1 - n = \text{odd}, \end{cases}$$

where

$$S_1(t_i) = \sum_{j=1}^{2n_1+2} \binom{2n_1+2}{j} (-1)^{j+1} G(t_{i-j}).$$

Since, by (11), $G(t_i) = g_i$ for $i = 1, 2, ..., 2n_1 + 1$, if we replace t_i by $-t_i$, then

$$S_1(-t_i) = \sum_{j=1}^{2n_1+2} {2n_1+2 \choose j} (-1)^{j+1} g_{-i+j}.$$

Hence we get the constraints at the left end as

$$\sum_{j=0}^{2n_1+2} \binom{2n_1+2}{j} (-1)^j g_{-i+j} = 0,$$
(31)

where

$$i = \begin{cases} 1, 2, \dots, \frac{4n_1 - n}{2}, & \text{if } 4n_1 - n = \text{even}, \\ 1, 2, \dots, \frac{4n_1 - n + 1}{2}, & \text{if } 4n_1 - n = \text{odd}. \end{cases}$$

4.2.2 The right end imposed constraints

For the right end imposed values, $g_i = S_1(t_i)$, i = N + 1, N + 2, ..., $N + (2n_1 - 1)$, and

$$S_1(t_i) = \sum_{j=1}^{2n_1+2} \binom{2n_1+2}{j} (-1)^{j+1} g_{i-j}$$

Hence we get the constraints at the right end as

$$\sum_{j=0}^{2n_1+2} \binom{2n_1+2}{j} (-1)^j g_{i-j} = 0,$$
(32)

where

$$i = \begin{cases} N+1, N+2, \dots, N+\frac{4n_1-n}{2}, & \text{if } 4n_1-n = \text{even}, \\ N+1, N+2, \dots, N+\frac{4n_1-n-1}{2}, & \text{if } 4n_1-n = \text{odd}. \end{cases}$$

5 The stable systems and their convergence

In this section, we present the linear and nonlinear stable systems of equations for the problems (3) and (4), respectively.

5.1 The linear stable system

Since the system (17) is unstable, by combining $4n_1 - n$ constraints, we get the stable system of the form

$$B_L G_L = D_L, \tag{33}$$

where the matrix B_L is defined as

$$B_L = \left((\mathbb{B}_L)^T, (A_L)^T, (\mathbb{B}_R)^T \right)^T, \tag{34}$$

 G_L is defined in (19), and D_L is the matrix of order $(N + 4n_1 + 1)$ defined as:

• If *n* is even,

$$D_{L} = (0, \dots, 0, 0, u^{n-1}(a), u^{n-2}(a), \dots, u'(a), u(a), d_{L}^{T}, u(b),$$
$$u'(b), \dots, u^{n-2}(b), u^{n-1}(b), 0, 0, \dots, 0)^{T};$$
(35)

• If *n* is odd,

$$\begin{aligned}
D_{L} &= (0, \dots, 0, 0, u^{n-2}(a), u^{n-3}(a), \dots, u'(a), u(a), d_{L}^{T}, u(b), \\
u'(b), \dots, u^{n-1}(b), 0, 0, \dots, 0)^{T}, \\
\text{or} \\
D_{L} &= (0, \dots, 0, 0, u^{n-1}(a), \dots, u'(a), u(a), d_{L}^{T}, u(b), \\
u'(b), \dots, u^{n-3}(b), u^{n-2}(b), 0, 0, \dots, 0)^{T},
\end{aligned}$$
(36)

where d_L is defined in (20). The matrix A_L is defined in (18), and the matrices \mathbb{B}_L and \mathbb{B}_R are of order $(\frac{4n_1}{2} \times (N + 4n_1 + 1))$ constructed as follows:

In matrix \mathbb{B}_L ,

- If $(4n_1 n)$ or n is even, the first $\frac{4n_1 n}{2}$ rows are constructed by using (31) for $i = \frac{4n_1 n}{2}, \frac{4n_1 n}{2} 1, \dots, 2, 1$, respectively. The last $\frac{n}{2}$ rows are obtained from $u^n(0), u^{n-1}(0), \dots, u'(0), u(0)$, respectively.
- If $(4n_1 n)$ or n is odd, the first $\frac{4n_1 n + 1}{2}$ rows are constructed by using (31) $i = \frac{4n_1 - n + 1}{2}, \frac{4n_1 - n + 1}{2} - 1, \dots, 2, 1$, respectively. The last $\frac{n}{2}$ rows are obtained from $u^n(0), u^{n-1}(0), \dots, u'(0), u(0)$, respectively.

The construction of \mathbb{B}_R is as follows:

- If $(4n_1 n)$ or *n* is even, then the first $\frac{n}{2}$ rows are obtained from $u(1), u'(1), \dots, u^{n-1}(1), u^n(1)$, respectively. The last $\frac{4n_1-n}{2}$ rows are constructed by using (32) for $i = N + 1, N + 2, \dots, N + \frac{4n_1-n}{2}$, respectively.
- If $(4n_1 n)$ or n is odd, then the first $\frac{n}{2}$ rows are obtained from $u(1), u'(1), \ldots, u^{n-1}(1), u^n(1)$, respectively. The last $\frac{4n_1-n-1}{2}$ rows are constructed by using (32) for $i = N + 1, N + 2, \ldots, N + \frac{4n_1-n-1}{2}$, respectively.

If derivative constraints are given, then first approximate them with the help of (29) and (30) before using them.

5.2 The nonlinear stable system

Since the system (25) is unstable, by combining $4n_1 - n$ imposed and n given constraints, we get a stable system of the form

$$B_{NL}G_{NL} = D_{NL}(g), \tag{37}$$

where the matrix B_{NL} is defined as

$$B_{NL} = \left((\mathbb{B}_L)^T, (A_{NL})^T, (\mathbb{B}_R)^T \right)^T,$$
(38)

 G_{NL} is defined in (27), and $D_{NL}(g)$ is the matrix of order $(N + 4n_1 + 1)$ defined as:

• If *n* is even,

$$D_{NL} = (0, \dots, 0, 0, u^{n-1}(a), u^{n-2}(a), \dots, u'(a), u(a), d_{NL}^{T}, u(b),$$
$$u'(b), \dots, u^{n-2}(b), u^{n-1}(b), 0, 0, \dots, 0)^{T};$$
(39)

• If *n* is odd,

$$D_{NL} = (0, ..., 0, 0, u^{n-2}(a), u^{n-3}(a), ..., u'(a), u(a), d_{NL}^{T}, u(b),$$

$$u'(b), ..., u^{n-1}(b), 0, 0, ..., 0)^{T},$$
or
$$D_{L} = (0, ..., 0, 0, u^{n-1}(a), ..., u'(a), u(a), d_{NL}^{T}, u(b),$$

$$u'(b), ..., u^{n-3}(b), u^{n-2}(b), 0, 0, ..., 0)^{T},$$
(40)

where d_{NL} is defined in (28). The matrix A_{NL} is defined in (26), and the matrices \mathbb{B}_L and \mathbb{B}_R of order $(\frac{4n_1}{2} \times (N + 4n_1 + 1))$ are same as in the case of a stable linear system.

5.3 Existence of the solution

The coefficient matrices B_L and B_{NL} of the linear and nonlinear stable systems are banded and nonsingular. Remember that these matrices are not symmetric or diagonally dominant, though it can be proved that B_L and B_{NL} are nonsingular/invertible. If we ignore the first and last few rows and columns then these are symmetric matrices. Now consider the square symmetric part of B_L and asymmetric matrix B_{NL} of order (N + 1) given by

It can be shown that if a(t) > 0 for $0 \le t \le 1$, B_L^1 and B_{NL}^2 are always nonsingular and, for large N, matrices B_L and B_{NL} are very similar to B_L^1 and B_{NL}^2 , respectively. Since these are banded matrices, by the results of Kilic and Stanica [9], their inverses exist by LU factorization.

5.4 The solutions of linear and nonlinear systems

Now, we discuss the methods to find the solutions of the systems (33) and (37).

5.4.1 The solution of linear system

The linear system of equations is defined in (33). We solve this system of equations by using Gaussian elimination method.

5.4.2 The solution of nonlinear system

For the solution of nonlinear system (37), we do a few steps: First of all, we solve the following linear system with initial approximation G_{NL}^0 :

$$B_{NL}G_{NL}^0 = D_{NL}^0,$$
 (41)

where

$$\begin{cases} D_{NL}^{0} = \{0, \dots, 0, 0, u^{n-1}(a), u^{n-2}(a), \dots, u'(a), u(a), F_{0}, F_{1}, \dots, F_{N}, u(b), \\ u'(b), \dots, u^{n-2}(b), u^{n-1}(b), 0, 0, \dots, 0\}, \\ F_{i} = h^{3}f(t_{i}, M_{i}, S), \quad i = 0, 1, 2, \dots, N, a \leq t \leq b, \\ M_{i} = u(a) + ih(\frac{u(b)-u(a)}{b-a}), \\ S = u(b) - u(a). \end{cases}$$

The solution of this system by Gaussian elimination method gives the initial approximation of the following nonlinear system:

$$B_{NL}G_{NL}^{m+1} = D_{NL}(G_{NL}^m), \quad m = 0, 1, 2, 3, \dots$$
(42)

Now continue the iterations by Gaussian elimination until

$$\left\|G_{NL}^{m} - G_{NL}^{m-1}\right\| \le tol\epsilon,\tag{43}$$

where *tole* is a chosen value. For example, someone can choose $tole = 10^{-6}$.

6 The special cases of our algorithms

Here we present several special cases of our algorithms. We see that the algorithms based on interpolating and approximating schemes for solving linear and nonlinear DEs with the set of constraints are special cases of our algorithms.

6.1 The special cases of our algorithms based on interpolating schemes

Here we see that a number of algorithms based on the refinement schemes for solving differential equations are the special cases of our algorithms.

- If we take n = 2, the problem (3) with (5) at a = 0 and b = 1 becomes a second-order linear DE. For its solution, if we put $n = n_1 = 2$ in (1), (2), (11), (31), (32), and (33) then we get the algorithms of Qu and Agarwal [2] and Mustafa et al. [10].
- If we take *n* = 2, the problem (4) with (5) at *a* = 0 and *b* = 1 becomes a second-order nonlinear DE. If we put *n* = *n*₁ = 2 in (1), (2), (11), (31), (32), and (37) then we get the algorithm of Qu and Agarwal [3].

- If we take n = 3, the problem (3) and (6) at a = 0 and b = 1 becomes a third-order linear DE. If we put $n = n_1 = 3$ in (1), (2), (11), (31), (32), and (33) then we get the algorithm of Mustafa and Ejaz [4, 11].
- If we take n = 3, the problem (4) and (6) at a = 0 and b = 1 becomes a third-order nonlinear DE. If we put $n = n_1 = 3$ in (1), (2), (11), (31), (32), and (37) then we get the algorithm of Ejaz and Mustafa [6].
- If we take n = 4, the problem (3) and (5) at a = 0 and b = 1 becomes a fourth-order linear DE. If we put $n = n_1 = 4$ in (1), (2), (11), (31), (32), and (33) then we get the algorithm of Ejaz et al. [5].
- If we take n = 4, the problem (4) and (5) at a = 0 and b = 1 becomes a fourth-order linear DE. If we put $n = n_1 = 4$ in (1), (2), (11), (31), (32), and (37) then we get the algorithms of Mustafa et al. [7] and Ejaz et al. [12].

6.2 The special cases of our algorithms based on approximating schemes

The algorithms presented in Sects. 3 and 4 are based on the interpolating refinement schemes. Such algorithms can be restructured to get the algorithms based on approximating refinement schemes.

For the achievement of the purpose, we first choose an appropriate C^n approximating refinement scheme with complexity $m_1 = 2, 4, 6, ...$ and which satisfies (10). Then we discuss its properties as we have done for interpolating refinement schemes which are given in (7), (8), and (9). Proposition 1 will be stated in a similar way for this case. The algorithms for the solutions of the problems (3)–(6) by approximating refinement schemes can be obtained by replacing $n_1 = \frac{m_1-2}{2}$ in the algorithms defined in Sects. 3 and 4.

Here we see that all the existing algorithms based on the approximating refinement schemes for solving differential equations will be special cases of our algorithms.

- If we take n = 2, the problem (3) and (5) at a = 0 and b = 1 becomes a second-order linear DE. Then select a suitable ($m_1 = 6$)-point approximating refinement scheme which satisfies (10). Further by substituting $m_1 = 6$ or $n_1 = 2$ in (11), (31), (32), and (33), we get the algorithm of Kanwal et al. [13].
- If we take n = 3, the problem (3) and (5) at a = 0 and b = 1 becomes a second-order linear DE. Then select a suitable (m₁ = 8)-point approximating refinement scheme which satisfies (10). Further by substituting m₁ = 8 or n₁ = 3 in (11), (31), (32), and (37), we get the algorithm of Manan et al. [14].
- If we take n = 4, the problem (3) and (5) at a = 0 and b = 1 becomes a fourth-order linear DE. Then select a suitable (m₁ = 10)-point approximating refinement scheme [15]. Further by substituting m₁ = 10 or n₄ = 4 in (11), (31), (32), and (33), we get the algorithm of Ejaz et al. [5].

7 Conclusion

In this paper, we first presented the generalized algorithms based on binary interpolating refinement schemes for the solution of the nth order linear and nonlinear differential equations with a set of constraints. Then we restructured these algorithms to get the algorithms based on approximating schemes. So, all the subdivision-based algorithms are easily restructured by substituting the suitable values of n and m in generalized algorithms for the solution of the nth order linear and nonlinear differential equations with a set of constraints. Hence, we showed that several algorithms based on the interpolating and approximating refinement schemes for solving differential equations are the special cases of our unified algorithms.

8 Limitations of generalized algorithms

In this section, we present limitations of our generalized algorithms based on binary interpolating refinement schemes for the solution of the *n*th order linear and nonlinear differential equations with a set of constraints.

- Our algorithm is a generalization of all the existing algorithms based on interpolating or approximating subdivision schemes.
- We can easily reconstruct an algorithm to find the solution of any order linear and nonlinear differential equations with a set of constraints just by substituting the suitable values of n (order of DEs) and m_1 (number of points in a subdivision scheme) in generalized algorithms.
- Our generalized algorithms reconstruct all the subdivision-based algorithms for the solution of the *n*th order linear and nonlinear ordinary differential equations with a set of constraints defined at a = 0 and b = 1.
- Our generalized algorithm is not applicable when the constraints are defined at points other than *a* = 0 and *b* = 1.

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Competing interests

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Authors' contributions

Conceptualization, GM and STE; Formal analysis, DB and Y-MC; Methodology, GM and DB; Supervision, GM; Writing original draft, STE and GM; Writing, reviewing, and editing, STE and Y-MC. All authors read and approved the final manuscript.

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