

Arbitrary-order trigonometric Fourier collocation methods for multi-frequency oscillatory systems

Bin Wang ^{a,b *}, Arieh Iserles ^c, and Xinyuan Wu ^{b **}

^aSchool of Mathematical Sciences, Qufu Normal University, Qufu, Shandong 273165, P.R.China.

^bDepartment of Mathematics, Nanjing University, State Key Laboratory for Novel Software Technology at Nanjing University, Nanjing 210093, P.R. China.

^cDepartment of Applied Mathematics and Theoretical Physics, Centre for Mathematical Sciences, University of Cambridge, Wilberforce Rd, Cambridge CB3 0WA, United Kingdom.

Communicated by Hans Munthe-Kaas

Abstract. We rigorously study a novel type of trigonometric Fourier collocation methods for solving multi-frequency oscillatory second-order ordinary differential equations (ODEs) $q''(t) + Mq(t) = f(q(t))$ with a principal frequency matrix $M \in \mathbb{R}^{d \times d}$. If M is symmetric and positive semi-definite and $f(q) = -\nabla U(q)$ for a smooth function $U(q)$, then this is a multi-frequency oscillatory Hamiltonian system with the Hamiltonian $H(q, p) = p^T p/2 + q^T M q/2 + U(q)$, where $p = q'$. The solution of this system is a nonlinear multi-frequency oscillator. The new trigonometric Fourier collocation method takes advantage of the special structure brought by the linear term Mq and its construction incorporates the idea of collocation methods, the variation-of-constants formula and the local Fourier expansion of the system. The properties of the new methods are analysed. The analysis in the paper demonstrates an important feature, namely that the trigonometric Fourier collocation methods can be of an arbitrary order and when $M \rightarrow 0$, each trigonometric Fourier collocation method creates a particular Runge–Kutta–Nyström-type Fourier collocation method, which is symplectic under some conditions. This allows us to obtain arbitrary high order symplectic methods to deal with a special and important class of systems of second-order ODEs in an efficient way. The results of numerical experiments are quite promising and show that the trigonometric Fourier collocation methods are significantly more efficient in comparison with alternative approaches that have previously appeared in the literature.

Keywords: Second-order ordinary differential equations · Multi-frequency oscillatory systems · Trigonometric Fourier collocation methods · Multi-

* Corresponding author, Department of Mathematics, Nanjing University, Nanjing 210093, P.R. China.

** E-mail: wangbinmaths@gmail.com (Bin Wang), A.Iserles@damtp.cam.ac.uk (Arieh Iserles), xywu@nju.edu.cn (Xinyuan Wu)

frequency oscillatory Hamiltonian systems · Quadratic invariant · Variation-of-constants formula · Symplectic methods

Mathematics Subject Classification 65L05 · 65L20 · 65M20 · 65P10

1 Introduction

In this paper we are concerned with systems of multi-frequency oscillatory second-order differential equations of the form

$$q''(t) + Mq(t) = f(q(t)), \quad q(0) = q_0, \quad q'(0) = q'_0, \quad t \in [0, t_{\text{end}}], \quad (1)$$

where M is a $d \times d$ positive semi-definite matrix implicitly containing the frequencies of the oscillatory problem and $f : \mathbb{R}^d \rightarrow \mathbb{R}^d$ is an analytic function. The solution of this system is a multi-frequency nonlinear oscillator because of the presence of the linear term Mq . (1) is a multi-frequency oscillatory system with multiple time scales and occurs in a wide variety of applications such as quantum physics, circuit simulations, flexible body dynamics, mechanics (see, e.g. [11, 12, 16, 21, 22, 57, 60]).

Once it is further assumed that M is symmetric and f is the negative gradient of a real-valued function $U(q)$, the system (1) is identical to the following initial value Hamiltonian system

$$\begin{aligned} \dot{q} &= \nabla_p H(q, p), & q(0) &= q_0, \\ \dot{p} &= -\nabla_q H(q, p), & p(0) &= p_0 \equiv q'_0 \end{aligned} \quad (2)$$

with the Hamiltonian function

$$H(q, p) = \frac{1}{2} p^T p + \frac{1}{2} q^T M q + U(q). \quad (3)$$

This is an important system which has been investigated by many authors (see, e.g. [9, 11, 12, 21, 22]). In essence, numerous mechanical systems with a partitioned Hamiltonian function fit this pattern. Two fundamental properties of Hamiltonian systems are:

(a) the solutions preserve the Hamiltonian H , i.e., $H(q(t), p(t)) \equiv H(q_0, p_0)$ for any $t \geq 0$;

(b) the corresponding flow is symplectic, i.e., it preserves the differential 2-form $\sum_{i=1}^d dp_i \wedge dq_i$.

An important feature of ordinary differential equations (ODEs) is first integrals. Much work deals with the conservation of invariants (first integrals) by numerical methods, e.g. [7, 13, 22–24, 33, 34, 39]. We consider the quadratic invariant $Q = q^T D p$ of (1). The quadratic form Q is a first integral of (1) if and only if $p^T D p + q^T D(f(q) - Mq) = 0$ for all $p, q \in \mathbb{R}^d$. This implies that D is a skew-symmetric matrix and that $q^T D(f(q) - Mq) = 0$ for any $q \in \mathbb{R}^d$. It has

been pointed out in [22, 24] that symplecticity is a quadratic first integral acting on the vector fields of a differential equation and every numerical method that preserves quadratic first integrals is a symplectic method for the corresponding Hamiltonian system.

For the Hamiltonian differential equations

$$y' = J^{-1}\nabla H(y) \quad (4)$$

with a skew-symmetric constant matrix J and the Hamiltonian $H(y)$, various numerical methods including symplectic methods and energy-preserving methods have been proposed and we refer the reader to [1, 3, 4, 6, 8, 10, 20, 23, 30, 31, 36, 39, 41, 43, 44, 46, 49, 51] for some related work. These methods are applicable to (2) since (2) can be rewritten in the form (4). However, it is noted that the system (1) or (2) is oscillatory due to the special structure brought by the linear term Mq . In the last few decades, the theory of numerical methods for oscillatory differential equations has reached a certain maturity. Many efficient codes, mainly based on Runge–Kutta–Nyström (RKN) methods and exponentially/trigonometrically-fitted methods, have become available. Among these various methods, exponential/trigonometric integrators are a very powerful approach and we refer to [28] for a survey work on exponential integrators. Other work related to exponential/trigonometric integrators can be found in [21, 22, 26, 27, 29, 51, 58] for instance. Much has been done to use the special structure brought by the term Mq to solve (1) effectively and we refer the reader to [14, 16–18, 21, 26, 37, 51–54, 56–61] for example.

The motivation for developing exponential/trigonometric algorithms for the oscillatory system (1) with the linear part Mq originates in a variety of research fields such as astronomy, quantum physics, molecular dynamics, mechanics, theoretical physics, semi-discrete wave equations approximated by the method of lines, and others from both applied and pure mathematics. It is apparent that the exponential/trigonometric methods taking advantage of the special structure brought by the term Mq not only produce an improved qualitative behaviour, but also allow for a more accurate long-term integration than with general-purpose methods. Inspired by this point, this paper is devoted to deriving and analysing a novel kind of trigonometric method (trigonometric Fourier collocation method) for (1).

When $M \rightarrow 0$, (1) reduces to a special and important class of systems of second-order ODEs expressed in the traditional form

$$q''(t) = f(q(t)), \quad q(0) = q_0, \quad q'(0) = q'_0, \quad t \in [0, t_{\text{end}}]. \quad (5)$$

Its corresponding Hamiltonian system and quadratic invariant can be described by the above statements for (1) with $M \rightarrow 0$. Therefore, we are hopeful of obtaining efficient methods for (5) based on the methods designed for (1) when $M \rightarrow 0$.

The main aim of this paper is to design a novel trigonometric Fourier collocation method of arbitrary order for solving multi-frequency oscillatory systems (1) and to obtain arbitrary order symplectic methods for system (5) based on

the new trigonometric Fourier collocation method. This *trigonometric Fourier collocation (TFC)* method is a kind of collocation method: collocation methods for ODEs (see, e.g. [20, 22, 24, 32, 55]) have a long history. An interesting feature of collocation methods is that we not only get a discrete set of approximations, but also a continuous approximation to the solution. The TFC methods not only share this feature of collocation methods, but also incorporate the special structure originating in the term Mq . Moreover, they are derived by truncating the local Fourier expansion of the system and thus it is very simple to get arbitrarily high order methods for solving (1). In addition, the TFC methods are applicable to the system (1) with no regard to symmetry of M . Besides, when $M \rightarrow 0$, our TFC method also sets out a special and efficient arbitrary order symplectic RKN approach to solving the special and important class of second-order ODEs (5). The fact stated above renders the TFC methods a very powerful and efficient approach in applications.

With this premise, this paper is organized as follows. Section 2 presents the local Fourier expansion of (1). The TFC methods are formulated in Section 3 and are analysed in Section 4. Section 5 is devoted to reporting several numerical experiments to demonstrate the efficiency of our novel approximation. Some conclusions are included in Section 6.

2 Local Fourier expansion of (1)

Before considering the Fourier expansion, we commence by recalling matrix-valued functions which have been first defined in [61]:

$$\phi_i(M) := \sum_{l=0}^{\infty} \frac{(-1)^l M^l}{(2l+i)!}, \quad i = 0, 1. \quad (6)$$

These functions reduce to the ϕ -functions used in Gautschi-type trigonometric or exponential integrators in [21, 22, 26] when M is a symmetric and positive semi-definite matrix. We note an important fact, namely that the convergent Taylor expansion in (6) depends directly on M and is applicable not only to symmetric matrices but also to nonsymmetric ones. By this definition and the formulation of our methods presented in next section, it can be observed that our methods are applicable to the system (1) with no regard to symmetry of M . This is the reason why we use the definition (6) in this paper.

Concerning the variation-of-constants formula for (1) given in [59], we have the following result on the exact solution of the system (1) and its derivative:

$$\begin{aligned} q(h) &= \phi_0(V)q_0 + h\phi_1(V)p_0 + h^2 \int_0^1 (1-z)\phi_1((1-z)^2V)f(q(hz))dz, \\ p(h) &= -hM\phi_1(V)q_0 + \phi_0(V)p_0 + h \int_0^1 \phi_0((1-z)^2V)f(q(hz))dz \end{aligned} \quad (7)$$

for $h > 0$, where $V = h^2 M$.

In what follows we consider the multi-frequency oscillatory system (1) restricted to the interval $[0, h]$:

$$q''(t) + Mq(t) = f(q(t)), \quad q(0) = q_0, \quad q'(0) = q'_0, \quad t \in [0, h]. \quad (8)$$

Choose an orthogonal polynomial basis $\{\widehat{P}_j\}_{j=0}^{\infty}$ on the interval $[0, 1]$: e.g., the shifted Legendre polynomials over the interval $[0, 1]$, scaled in order to be orthonormal. Consequently,

$$\int_0^1 \widehat{P}_i(x)\widehat{P}_j(x)dx = \delta_{ij}, \quad \deg(\widehat{P}_j) = j, \quad i, j \geq 0,$$

where δ_{ij} is the Kronecker symbol. We rewrite the right-hand side of (8) as

$$f(q(\xi h)) = \sum_{j=0}^{\infty} \widehat{P}_j(\xi)\gamma_j(q), \quad \xi \in [0, 1]; \quad \gamma_j(q) := \int_0^1 \widehat{P}_j(\tau)f(q(\tau h))d\tau. \quad (9)$$

(To simplify the notation, we use $\gamma_j(q)$ to denote the coefficients involved in the Fourier expansion, replacing the more complete notation $\gamma_j(h, f(q))$.)

Combining (7) and (9), we immediately have the following result.

Theorem 21 *The solution of (8) and its derivative satisfy*

$$\begin{aligned} q(h) &= \phi_0(V)q_0 + h\phi_1(V)p_0 + h^2 \sum_{j=0}^{\infty} I_{1,j}(V)\gamma_j(q), \\ p(h) &= -hM\phi_1(V)q_0 + \phi_0(V)p_0 + h \sum_{j=0}^{\infty} I_{2,j}(V)\gamma_j(q), \end{aligned} \quad (10)$$

where

$$I_{1,j}(V) := \int_0^1 \widehat{P}_j(z)(1-z)\phi_1((1-z)^2V)dz, \quad I_{2,j}(V) := \int_0^1 \widehat{P}_j(z)\phi_0((1-z)^2V)dz. \quad (11)$$

In what follows we consider the truncated Fourier expansion appeared first in [3]. The main point in designing practical schemes to solve (1) is based on truncating the series (9) after $r \geq 2$ terms and this replaces the initial value problem (1) by the approximate problem

$$\begin{aligned} \tilde{q}'(\xi h) &= \tilde{p}(\xi h), & \tilde{q}(0) &= q_0, \\ \tilde{p}'(\xi h) &= -M\tilde{q}(\xi h) + \sum_{j=0}^{r-1} \widehat{P}_j(\xi)\gamma_j(\tilde{q}), & \tilde{p}(0) &= p_0, \end{aligned}$$

which is a natural extension of the formulae presented in [2].

The implicit solution of the new problem is

$$\begin{aligned}\tilde{q}(h) &= \phi_0(V)q_0 + h\phi_1(V)p_0 + h^2 \sum_{j=0}^{r-1} I_{1,j}(V)\gamma_j(\tilde{q}), \\ \tilde{p}(h) &= -hM\phi_1(V)q_0 + \phi_0(V)p_0 + h \sum_{j=0}^{r-1} I_{2,j}(V)\gamma_j(\tilde{q}).\end{aligned}\tag{12}$$

3 Formulation of the TFC methods

It is clear that the scheme (12) itself falls well short of being a practical method unless the integrals $I_{1,j}(V)$, $I_{2,j}(V)$, $\gamma_j(\tilde{q})$ in (12) can be approximated.

3.1 The computation of $I_{1,j}(V)$, $I_{2,j}(V)$

According to the definition of shifted Legendre polynomials on the interval $[0, 1]$:

$$\widehat{P}_j(x) = (-1)^j \sqrt{2j+1} \sum_{k=0}^j \binom{j}{k} \binom{j+k}{k} (-x)^k, \quad j = 0, 1, \dots, \quad x \in [0, 1],\tag{13}$$

we arrive at

$$\begin{aligned}I_{1,j}(V) &= \int_0^1 \widehat{P}_j(z)(1-z)\phi_1((1-z)^2V)dz \\ &= \sqrt{2j+1} \sum_{l=0}^{\infty} (-1)^j \sum_{k=0}^j \binom{j}{k} \binom{j+k}{k} \int_0^1 (-z)^k (1-z)^{2l+1} dz \frac{(-1)^l V^l}{(2l+1)!} \\ &= \sqrt{2j+1} \sum_{l=0}^{\infty} \sum_{k=0}^j (-1)^{j+k} \binom{j}{k} \binom{j+k}{k} \frac{k!(2l+1)!}{(2l+k+2)!} \frac{(-1)^l V^l}{(2l+1)!} \\ &= \sqrt{2j+1} \sum_{l=0}^{\infty} \sum_{k=0}^j \frac{(-1)^{j+k+l} (j+k)!}{k!(j-k)!(2l+k+2)!} V^l, \quad j = 0, 1, \dots\end{aligned}$$

Recall that a *generalized hypergeometric function* (see [42, 48]) is

$${}_mF_n \left[\begin{matrix} \alpha_1, \alpha_2, \dots, \alpha_m; \\ \beta_1, \beta_2, \dots, \beta_n; \end{matrix} x \right] = \sum_{l=0}^{\infty} \frac{\prod_{i=1}^m (\alpha_i)_l}{\prod_{i=1}^n (\beta_i)_l} \frac{x^l}{l!},\tag{14}$$

where the *Pochhammer symbol* $(z)_l$ is defined as $(z)_0 = 1$ and $(z)_l = z(z+1)\cdots(z+l-1)$, $l \in \mathbb{N}$. The parameters α_i and β_i are arbitrary complex numbers, except that β_i can be neither zero nor a negative integer. $I_{1,j}(V)$ can be

formulated as

$$\begin{aligned} I_{1,j}(V) &= \sqrt{2j+1} \sum_{l=0}^{\infty} \frac{(-1)^{j+l}}{(2l+2)!} {}_2F_1 \left[\begin{matrix} -j, j+1; \\ 2l+3; \end{matrix} \middle| 1 \right] V^l \\ &= \sqrt{2j+1} \sum_{l=\lfloor j/2 \rfloor}^{\infty} \frac{(-1)^{j+l} (2l+1)!}{(2l+2+j)!(2l+1-j)!} V^l, \end{aligned}$$

where we have used the standard formula to sum up ${}_2F_1$ functions with unit argument (see [42]).

For $n \geq 0$, $(2l+n)! = n! 2^{2l} \binom{n+1}{2}_l \binom{n+2}{2}_l$ and from this, it follows that

$$\begin{aligned} I_{1,2j}(V) &= \sqrt{4j+1} \sum_{l=j}^{\infty} \frac{(-1)^{2j+l} (2l+1)!}{(2l+2+2j)!(2l+1-2j)!} V^l \\ &= \sqrt{4j+1} (-1)^j V^j \sum_{l=0}^{\infty} \frac{(-1)^l (2l+2j+1)!}{(2l+4j+2)!(2l+1)!} V^l \\ &= \sqrt{4j+1} (-1)^j V^j \sum_{l=0}^{\infty} \frac{(2j+1)!(j+1)_l (j+\frac{3}{2})_l}{(4j+2)!(2j+\frac{3}{2})_l (2j+2)_l (\frac{3}{2})_l} \frac{(-V/4)^l}{l!} \\ &= (-1)^j \sqrt{4j+1} \frac{(2j+1)!}{(4j+2)!} V^j {}_2F_3 \left[\begin{matrix} j+1, j+\frac{3}{2}; \\ 2j+2, \frac{3}{2}, 2j+\frac{3}{2}; \end{matrix} \middle| -\frac{V}{4} \right], \quad j = 0, 1, \dots \end{aligned} \tag{15}$$

$I_{1,2j+1}(V)$ can be dealt with similarly:

$$I_{1,2j+1}(V) = (-1)^{j+1} \sqrt{4j+3} \frac{(2j+1)!}{(4j+3)!} V^j {}_2F_3 \left[\begin{matrix} j+1, j+\frac{3}{2}; \\ 2j+2, \frac{1}{2}, 2j+\frac{5}{2}; \end{matrix} \middle| -\frac{V}{4} \right], \quad j = 0, 1, \dots \tag{16}$$

Likewise, we can obtain $I_{2,j}(V)$:

$$\begin{aligned} I_{2,2j}(V) &= (-1)^j \sqrt{4j+1} \frac{(2j)!}{(4j+1)!} V^j {}_2F_3 \left[\begin{matrix} j+\frac{1}{2}, j+1; \\ 2j+1, \frac{1}{2}, 2j+\frac{3}{2}; \end{matrix} \middle| -\frac{V}{4} \right], \\ I_{2,2j+1}(V) &= (-1)^j \sqrt{4j+3} \frac{(2j+2)!}{(4j+4)!} V^{j+1} {}_2F_3 \left[\begin{matrix} j+\frac{3}{2}, j+2; \\ 2j+3, \frac{3}{2}, 2j+\frac{5}{2}; \end{matrix} \middle| -\frac{V}{4} \right], \quad j = 0, 1, \dots \end{aligned} \tag{17}$$

Based on the formula from MathWorld—A Wolfram Web Resource

$${}_2F_3 \left[\begin{matrix} a, a+1/2; \\ 2a, d, 2a-d+1; \end{matrix} \middle| z \right] = {}_0F_1 \left[\begin{matrix} -; z \\ d; \end{matrix} \middle| \frac{z}{4} \right] {}_0F_1 \left[\begin{matrix} -; \\ 2a-d+1; \end{matrix} \middle| \frac{z}{4} \right],$$

the above results can be simplified as

$$\begin{aligned}
I_{1,2j}(V) &= (-1)^j \sqrt{4j+1} \frac{(2j+1)!}{(4j+2)!} V^j {}_0F_1 \left[\begin{matrix} - \\ \frac{3}{2} \end{matrix}; -\frac{V}{16} \right] {}_0F_1 \left[\begin{matrix} - \\ 2j + \frac{3}{2} \end{matrix}; -\frac{V}{16} \right], \\
I_{1,2j+1}(V) &= (-1)^{j+1} \sqrt{4j+3} \frac{(2j+1)!}{(4j+3)!} V^j {}_0F_1 \left[\begin{matrix} - \\ \frac{1}{2} \end{matrix}; -\frac{V}{16} \right] {}_0F_1 \left[\begin{matrix} - \\ 2j + \frac{5}{2} \end{matrix}; -\frac{V}{16} \right], \\
I_{2,2j}(V) &= (-1)^j \sqrt{4j+1} \frac{(2j)!}{(4j+1)!} V^j {}_0F_1 \left[\begin{matrix} - \\ \frac{1}{2} \end{matrix}; -\frac{V}{16} \right] {}_0F_1 \left[\begin{matrix} - \\ 2j + \frac{3}{2} \end{matrix}; -\frac{V}{16} \right], \\
I_{2,2j+1}(V) &= (-1)^j \sqrt{4j+3} \frac{(2j+2)!}{(4j+4)!} V^{j+1} {}_0F_1 \left[\begin{matrix} - \\ \frac{3}{2} \end{matrix}; -\frac{V}{16} \right] {}_0F_1 \left[\begin{matrix} - \\ 2j + \frac{5}{2} \end{matrix}; -\frac{V}{16} \right], \quad j = 0, 1, \dots
\end{aligned} \tag{18}$$

It should be noted that $\phi_0(V)$ and $\phi_1(V)$ can also be expressed by the generalized hypergeometric function ${}_0F_1$:

$$\phi_0(V) = {}_0F_1 \left[\begin{matrix} - \\ \frac{1}{2} \end{matrix}; -\frac{V}{4} \right], \quad \phi_1(V) = {}_0F_1 \left[\begin{matrix} - \\ \frac{3}{2} \end{matrix}; -\frac{V}{4} \right]. \tag{19}$$

A Bessel function of the first kind can be expressed in terms of function ${}_0F_1$ by (see [40])

$$J_n(x) = \frac{(x/2)^n}{n!} {}_0F_1 \left[\begin{matrix} - \\ n+1 \end{matrix}; -\frac{x^2}{4} \right].$$

We also note that ${}_0F_1 \left[\begin{matrix} - \\ \frac{n}{2} \end{matrix}; -S \right]$ with $n \in \mathbb{N}$ and any symmetric matrix S also is viewed as a Bessel function of matrix argument (see [45]). The importance of this hypergeometric representation is that most modern software, e.g., Maple, Mathematica, and Matlab, is well equipped to calculate generalized hypergeometric functions. There are also much work concerning the evaluation of generalized hypergeometric functions of a matrix argument, see, e.g. [5, 19, 35, 45].

3.2 Discretization

The key question now is how to deal with $\gamma_j(\tilde{q})$. This can be done by introducing a quadrature formula based at k ($k \geq r$) abscissae $0 \leq c_1 \leq \dots \leq c_k \leq 1$, thus obtaining an approximation of the form

$$\gamma_j(\tilde{q}) \approx \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(\tilde{q}(c_l h)), \quad j = 0, 1, \dots, r-1, \tag{20}$$

where b_l , $l = 1, 2, \dots, k$, are the quadrature weights. Based on (20) and (12), it is natural to consider the following scheme

$$\begin{aligned}
v(h) &= \phi_0(V)q_0 + h\phi_1(V)p_0 + h^2 \sum_{j=0}^{r-1} I_{1,j}(V) \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v(c_l h)), \\
u(h) &= -hM\phi_1(V)q_0 + \phi_0(V)p_0 + h \sum_{j=0}^{r-1} I_{2,j}(V) \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v(c_l h)),
\end{aligned}$$

which is the exact solution of the oscillatory initial value problem

$$\begin{aligned} v'(\xi h) &= u(\xi h), & v(0) &= q_0, \\ u'(\xi h) &= -Mv(\xi h) + \sum_{j=0}^{r-1} \widehat{P}_j(\xi) \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v(c_l h)), & u(0) &= p_0. \end{aligned} \quad (21)$$

From (21), it follows that $v(c_i h)$, $i = 1, 2, \dots, k$, can be obtained by solving the following discrete problems:

$$v''(c_i h) + Mv(c_i h) = \sum_{j=0}^{r-1} \widehat{P}_j(c_i) \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v(c_l h)), \quad v(0) = q_0, \quad v'(0) = p_0. \quad (22)$$

By setting, as usual, $v_i = v(c_i h)$, $i = 1, 2, \dots, k$, (22) can be solved by the variation-of-constants formula in the form:

$$\begin{aligned} v_i &= \phi_0(c_i^2 V) q_0 + c_i h \phi_1(c_i^2 V) p_0 \\ &\quad + (c_i h)^2 \sum_{j=0}^{r-1} \int_0^1 \widehat{P}_j(c_i z) (1-z) \phi_1((1-z)^2 c_i^2 V) dz \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v_l), \\ & \quad i = 1, 2, \dots, k. \end{aligned}$$

3.3 The TFC methods

We are now in a position to present our TFC methods for the multi-frequency oscillatory second-order ODEs (1) or the multi-frequency oscillatory Hamiltonian system (2).

Definition 31 *A trigonometric Fourier collocation (TFC) method for integrating the multi-frequency oscillatory system (1) or (2) is defined as*

$$\begin{aligned} v_i &= \phi_0(c_i^2 V) q_0 + c_i h \phi_1(c_i^2 V) p_0 + (c_i h)^2 \sum_{j=0}^{r-1} I_{1,j,c_i}(V) \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v_l), \\ & \quad i = 1, 2, \dots, k, \\ v(h) &= \phi_0(V) q_0 + h \phi_1(V) p_0 + h^2 \sum_{j=0}^{r-1} I_{1,j}(V) \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v_l), \\ u(h) &= -h M \phi_1(V) q_0 + \phi_0(V) p_0 + h \sum_{j=0}^{r-1} I_{2,j}(V) \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v_l), \end{aligned} \quad (23)$$

where h is the stepsize, \widehat{P}_j are defined by (13) and c_l , b_l , $l = 1, 2, \dots, k$ are the node points and the quadrature weights of a quadrature formula, respectively. Here r is an integer with the requirement: $2 \leq r \leq k$. $I_{1,j}(V)$, $I_{2,j}(V)$ can be computed by (15)–(17) and the $I_{1,j,c_i}(V)$ are defined as

$$\begin{aligned}
I_{1,j,c_i}(V) &:= \int_0^1 \widehat{P}_j(c_i z)(1-z)\phi_1((1-z)^2 c_i^2 V) dz \\
&= \sqrt{2j+1} \sum_{l=0}^{\infty} \frac{(-1)^{j+l}}{(2l+2)!} {}_2F_1 \left[\begin{matrix} -j, j+1; \\ 2l+3; \end{matrix} c_i \right] (c_i^2 V)^l \\
&= (-1)^j \frac{\sqrt{2j+1}(1-c_i)}{c_i} \sum_{l=0}^{\infty} P_j^{-2l-2}(1-2c_i)(c_i(c_i-1)V)^l,
\end{aligned}$$

where $P_j^{-2l-2}(1-2c_i)$ are associated Legendre functions (see [47]).

Remark 32 The TFC method (23) approximates the solution $q(t)$, $p(t)$ of the system (2) in the time interval $[0, h]$. Obviously, the values $v(h)$, $u(h)$ can be considered as the initial condition for a new initial value problem approximating $q(t)$, $p(t)$ in the time interval $[h, 2h]$. In general, one can extend the TFC methods in the usual time-stepping manner to the interval $[(i-1)h, ih]$ for any $i \geq 2$ and finally achieve a TFC method for $q(t)$, $p(t)$ in an arbitrary interval $[0, Nh]$.

Remark 33 It can be observed clearly from (23) that our TFC methods do not need the symmetry of M . Therefore, the method (23) can be applied to the system (1) with an arbitrary matrix M . Moreover, the method (23) integrates exactly the linear system $q'' + Mq = 0$. Thus it has an additional advantage of energy preservation and quadratic invariant preservation for the linear system.

Based on (23), we consider Fourier collocation methods for integrating a special and important class of systems of second-order ODEs (5). When $M \rightarrow 0$, $I_{1,j}(V)$ and $I_{2,j}(V)$ in (23) become

$$\begin{aligned}
\tilde{I}_{1,j} &:= I_{1,j}(0) = \int_0^1 \widehat{P}_j(z)(1-z)\phi_1(0) dz = (-1)^j \frac{\sqrt{2j+1}}{2} {}_2F_1 \left[\begin{matrix} -j, j+1; \\ 3; \end{matrix} 1 \right], \\
\tilde{I}_{2,j} &:= I_{2,j}(0) = \int_0^1 \widehat{P}_j(z)\phi_0(0) dz = (-1)^j \sqrt{2j+1} {}_2F_1 \left[\begin{matrix} -j, j+1; \\ 2; \end{matrix} 1 \right].
\end{aligned}$$

In order to simplify the results of $\sum_{j=0}^{r-1} \tilde{I}_{1,j} \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v_l)$ and $\sum_{j=0}^{r-1} \tilde{I}_{2,j} \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v_l)$ with $r \geq 2$, we compute

$$\tilde{I}_{1,j} = \begin{cases} \frac{1}{2}, & j = 0, \\ \frac{-1}{2\sqrt{3}}, & j = 1, \\ 0, & j \geq 2, \end{cases} \quad \tilde{I}_{2,j} = \begin{cases} 1, & j = 0, \\ 0, & j \geq 1. \end{cases}$$

Then we have

$$\begin{aligned} \sum_{j=0}^{r-1} \tilde{I}_{1,j} \sum_{l=1}^k b_l \hat{P}_j(c_l) f(v_l) &= \sum_{l=1}^k (1 - c_l) b_l f(v_l), \\ \sum_{j=0}^{r-1} \tilde{I}_{1,j} \sum_{l=1}^k b_l \hat{P}_j(c_l) f(v_l) &= \sum_{l=1}^k b_l f(v_l). \end{aligned}$$

All this can be summed up in the following definition.

Definition 34 An RKN-type Fourier collocation method for integrating the system (5) is defined as

$$\begin{aligned} v_i &= q_0 + c_i h p_0 + (c_i h)^2 \sum_{j=0}^{r-1} \tilde{I}_{1,j,c_i} \sum_{l=1}^k b_l \hat{P}_j(c_l) f(v_l), \quad i = 1, 2, \dots, k, \\ v(h) &= q_0 + h p_0 + h^2 \sum_{l=1}^k (1 - c_l) b_l f(v_l), \\ u(h) &= p_0 + h \sum_{l=1}^k b_l f(v_l), \end{aligned} \tag{24}$$

where h is the stepsize, c_l , b_l , $l = 1, 2, \dots, k$, are the node points and the quadrature weights of a quadrature formula, respectively, and \tilde{I}_{1,j,c_i} are defined by

$$\tilde{I}_{1,j,c_i} := (-1)^j \frac{\sqrt{2j+1}}{2} {}_2F_1 \left[\begin{matrix} -j, j+1 \\ 3 \end{matrix}; c_i \right] = (-1)^j \frac{\sqrt{2j+1}(1-c_i)}{c_i} P_j^{-2}(1-2c_i). \tag{25}$$

Remark 35 It can be observed that the method (24) is the subclass of k -stage RKN methods with the following Butcher tableau:

$$\begin{array}{c|c} c_1 & c_1^2 b_1 \sum_{j=0}^{r-1} \tilde{I}_{1,j,c_1} \hat{P}_j(c_1) \dots c_1^2 b_k \sum_{j=0}^{r-1} \tilde{I}_{1,j,c_1} \hat{P}_j(c_k) \\ \vdots & \vdots \quad \quad \quad \ddots \quad \quad \quad \vdots \\ c_k & c_k^2 b_1 \sum_{j=0}^{r-1} \tilde{I}_{1,j,c_k} \hat{P}_j(c_1) \dots c_k^2 b_k \sum_{j=0}^{r-1} \tilde{I}_{1,j,c_k} \hat{P}_j(c_k) \\ \hline \bar{a} & (1-c_1)b_1 \quad \quad \quad \dots \quad \quad \quad (1-c_k)b_k \\ \bar{b}^T & b_1 \quad \quad \quad \dots \quad \quad \quad b_k \\ \hline b^T & \end{array} =$$

From the analysis given in next section, it follows that the method (24) can attain arbitrary algebraic order and is symplectic when choosing a k -point Gauss-Legendre's quadrature and $r = k$. This feature is significant for solving the traditional second-order ODEs (5). The point stated above demonstrates the wider applications of the new TFC methods (23) and it makes our methods more efficient and competitive. Problem 4 in Section 5 will demonstrate this point clearly.

4 Properties of the TFC methods

In this section we first analyse the degree of accuracy of the TFC methods in preserving the solution $p(t)$, $q(t)$, the quadratic invariant Q and the Hamiltonian H . Other properties including convergence and stability properties are studied as well.

The following result is needed in our analysis and its proof can be found in [3].

Lemma 41 *Let $g : [0, h] \rightarrow \mathbb{R}^d$ have j continuous derivatives in the interval $[0, h]$. Then $\int_0^1 \widehat{P}_j(\tau)g(\tau h)d\tau = \mathcal{O}(h^j)$. As a consequence, $\gamma_j(v) = \mathcal{O}(h^j)$.*

Let the k -point quadrature formula in (23) have order $m - 1$, i.e., be exact for polynomials of degree less than or equal to $m - 1$ ($m \geq k$). Then we have

$$\Delta_j(h) := \gamma_j(v) - \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v_l) = \mathcal{O}(h^{m-j}), \quad j = 0, 1, \dots, r - 1.$$

Clearly, since $k \geq r$ is assumed, we obtain $m \geq r$ and this guarantees that the above $\Delta_j(h)$ can have good accuracy for any $j = 0, 1, \dots, r - 1$. Choosing k large enough, along with a suitable choice of the b_l and c_l , allows us to approximate the given integral $\gamma_j(v)$ to any degree of accuracy.

For the exact solution of (2), let $\mathbf{y}(h) = \left(q^T(h), p^T(h) \right)^T$. Then the oscillatory Hamiltonian system (2) can be rewritten in the form

$$\mathbf{y}'(\xi h) = F(\mathbf{y}(\xi h)) := \begin{pmatrix} p(\xi h) \\ -Mq(\xi h) + \sum_{j=0}^{\infty} \widehat{P}_j(\xi) \gamma_j(q) \end{pmatrix}, \quad \mathbf{y}_0 = \begin{pmatrix} q_0 \\ p_0 \end{pmatrix}. \quad (26)$$

The Hamiltonian is

$$H(\mathbf{y}) = \frac{1}{2} p^T p + \frac{1}{2} q^T M q + U(q). \quad (27)$$

On the other hand, for the TFC method (23), denoting $\omega(h) = \left(v^T(h), u^T(h) \right)^T$, the numerical solution satisfies

$$\omega'(\xi h) = \begin{pmatrix} u(\xi h) \\ -Mv(\xi h) + \sum_{j=0}^{r-1} \widehat{P}_j(\xi) \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v(c_l h)) \end{pmatrix}, \quad \omega_0 = \begin{pmatrix} q_0 \\ p_0 \end{pmatrix}. \quad (28)$$

4.1 The order

To express the dependence of the solutions of $\mathbf{y}'(t) = F(\mathbf{y}(t))$ on the initial values, for any given $\tilde{t} \in [0, h]$, we denote by $\mathbf{y}(\cdot, \tilde{t}, \tilde{\mathbf{y}})$ the solution satisfying the initial condition $\mathbf{y}(\tilde{t}, \tilde{t}, \tilde{\mathbf{y}}) = \tilde{\mathbf{y}}$ and set

$$\Phi(s, \tilde{t}, \tilde{\mathbf{y}}) = \frac{\partial \mathbf{y}(s, \tilde{t}, \tilde{\mathbf{y}})}{\partial \tilde{\mathbf{y}}}. \quad (29)$$

Recalling the elementary theory of ODEs, we have the following standard result (see, e.g. [25])

$$\frac{\partial \mathbf{y}(s, \tilde{t}, \tilde{\mathbf{y}})}{\partial \tilde{t}} = -\Phi(s, \tilde{t}, \tilde{\mathbf{y}})F(\tilde{\mathbf{y}}). \quad (30)$$

The following theorem states the result on the order of our TFC methods.

Theorem 42 *Let the quadrature formula in (23) have order $m - 1$. Then for the TFC method (23)*

$$\mathbf{y}(h) - \omega(h) = \mathcal{O}(h^{n+1}) \text{ with } n = \min\{m, 2r\}.$$

The RKN-type Fourier collocation method (24) is also of order n .

Proof. It follows from Lemma 41, (29) and (30) that

$$\begin{aligned} \mathbf{y}(h) - \omega(h) &= \mathbf{y}(h, 0, \mathbf{y}_0) - \mathbf{y}(h, h, \omega(h)) = - \int_0^h \frac{d\mathbf{y}(h, \tau, \omega(\tau))}{d\tau} d\tau \\ &= - \int_0^h \left[\frac{\partial \mathbf{y}(h, \tau, \omega(\tau))}{\partial \tilde{t}} + \frac{\partial \mathbf{y}(h, \tau, \omega(\tau))}{\partial \tilde{\mathbf{y}}} \omega'(\tau) \right] d\tau \\ &= h \int_0^1 \Phi(h, \xi h, \omega(\xi h)) \left[F(\omega(\xi h)) - \omega'(\xi h) \right] d\xi \\ &= h \int_0^1 \Phi(h, \xi h, \omega(\xi h)) \left(\begin{array}{c} \mathbf{0} \\ \sum_{j=0}^{r-1} \hat{P}_j(\xi) \Delta_j(h) + \sum_{j=r}^{\infty} \hat{P}_j(\xi) \gamma_j(v) \end{array} \right) d\xi. \end{aligned}$$

We rewrite $\Phi(h, \xi h, \omega(\xi h))$ as a block matrix,

$$\Phi(h, \xi h, \omega(\xi h)) = \begin{pmatrix} \Phi_{11}(\xi h) & \Phi_{12}(\xi h) \\ \Phi_{21}(\xi h) & \Phi_{22}(\xi h) \end{pmatrix},$$

where Φ_{ij} ($i, j = 1, 2$) are all $d \times d$ matrices. Then we obtain

$$\begin{aligned} \mathbf{y}(h) - \omega(h) &= h \begin{pmatrix} \int_0^1 \Phi_{12}(\xi h) \left(\sum_{j=0}^{r-1} \hat{P}_j(\xi) \Delta_j(h) + \sum_{j=r}^{\infty} \hat{P}_j(\xi) \gamma_j(v) \right) d\xi \\ \int_0^1 \Phi_{22}(\xi h) \left(\sum_{j=0}^{r-1} \hat{P}_j(\xi) \Delta_j(h) + \sum_{j=r}^{\infty} \hat{P}_j(\xi) \gamma_j(v) \right) d\xi \end{pmatrix} \\ &= h \begin{pmatrix} \sum_{j=0}^{r-1} \int_0^1 \Phi_{12}(\xi h) \hat{P}_j(\xi) d\xi \Delta_j(h) + \sum_{j=r}^{\infty} \int_0^1 \Phi_{12}(\xi h) \hat{P}_j(\xi) d\xi \gamma_j(v) \\ \sum_{j=0}^{r-1} \int_0^1 \Phi_{22}(\xi h) \hat{P}_j(\xi) d\xi \Delta_j(h) + \sum_{j=r}^{\infty} \int_0^1 \Phi_{22}(\xi h) \hat{P}_j(\xi) d\xi \gamma_j(v) \end{pmatrix} \\ &= h \left(\sum_{j=0}^{r-1} \mathcal{O}(h^j \times h^{m-j}) + \sum_{j=r}^{\infty} \mathcal{O}(h^j \times h^j) \right) = \mathcal{O}(h^{m+1}) + \mathcal{O}(h^{2r+1}), \end{aligned}$$

which completes the proof. \square

4.2 The order of energy preservation and quadratic invariant preservation

In what follows we are concerned with the order of preserving the Hamiltonian energy by our TFC method.

Theorem 43 *Under the condition in Theorem 42, we have*

$$H(\omega(h)) - H(\mathbf{y}_0) = \mathcal{O}(h^{n+1}) \text{ with } n = \min\{m, 2r\}.$$

The RKN-type Fourier collocation method (24) has the same order in preserving the Hamiltonian energy as the TFC methods.

We then consider the quadratic invariant $Q(\mathbf{y}) = q^T Dp$ of (1). The following result states the degree of accuracy of the TFC method (23) in its preservation.

Theorem 44 *Under the condition in Theorem 42, we have*

$$Q(\omega(h)) - Q(\mathbf{y}_0) = \mathcal{O}(h^{n+1}) \text{ with } n = \min\{m, 2r\}.$$

The proofs of these two theorems are given in Appendix A and Appendix B, respectively. It is well-known that n th order numerical methods can preserve the Hamiltonian energy or the quadratic invariant with at least n th degree of accuracy but unfortunately it follows from the proofs of these two theorems that our methods preserve the Hamiltonian energy and the quadratic invariant only with n th degree of accuracy. However, we prove in the next theorem that when $M \rightarrow 0$, the Fourier collocation methods can be symplectic, i.e., they exactly preserve the quadratic invariant.

A better result in preserving the quadratic invariant can be obtained for the RKN-type Fourier collocation method (24) and we state it in the following theorem.

Theorem 45 *Under the condition that c_l , $l = 1, 2, \dots, k$ are chosen as the node points of a k -point Gauss–Legendre’s quadrature over the integral $[0, 1]$ and $r = k$, the RKN-type Fourier collocation method (24) is symplectic, i.e., it preserves the quadratic invariant of (5) exactly.*

Proof. For any $m \in (1, 2, \dots, k)$, let

$$h(x) := x^2 \sum_{j=0}^{k-1} \widehat{P}_j(c_m) \tilde{I}_{1,j,x} - c_m^2 \sum_{j=0}^{k-1} \tilde{I}_{1,j,c_m} \widehat{P}_j(x) - x + c_m.$$

By the formulas (13), (25) and the fact that c_l are the node points of a Gauss–Legendre’s quadrature, it is true that

$$h(x) = (ax + b) \widehat{P}_k(x),$$

where

$$a = \frac{(-1)^k}{\sqrt{2k+1}} \left[c_m^2 \sum_{j=0}^{k-1} (-1)^j \sqrt{2j+1} \tilde{I}_{1,j,c_m} (j^2 + j - k^2 - k) + k(k+1)c_m - 1 \right],$$

$$b = \frac{(-1)^k}{\sqrt{2k+1}} \left[c_m - c_m^2 \sum_{j=0}^{k-1} (-1)^j \sqrt{2j+1} \tilde{I}_{1,j,c_m} \right].$$

From the fact that $\hat{P}_k(c_n) \equiv 0$, $n = 1, 2, \dots, k$, it follows that $h(c_n) \equiv 0$ and thus

$$c_m^2 \sum_{j=0}^{k-1} \tilde{I}_{1,j,c_m} \hat{P}_j(c_n) - c_n^2 \sum_{j=0}^{k-1} \tilde{I}_{1,j,c_n} \hat{P}_j(c_m) = c_m - c_n.$$

Therefore, we have

$$\begin{aligned} & b_m(\bar{b}_n - \bar{a}_{mn}) - b_n(\bar{b}_m - \bar{a}_{nm}) = b_m((1 - c_n)b_n - \bar{a}_{mn}) - b_n((1 - c_m)b_m - \bar{a}_{nm}) \\ & = b_m b_n \left(c_m - c_n - c_m^2 \sum_{j=0}^{k-1} \tilde{I}_{1,j,c_m} \hat{P}_j(c_n) + c_n^2 \sum_{j=0}^{k-1} \tilde{I}_{1,j,c_n} \hat{P}_j(c_m) \right) = b_m b_n 0 = 0. \end{aligned}$$

The symplecticity conditions of RKN methods (see [22, 24]) ensure the conclusion immediately. \square

Remark 46 *This result means that choosing a suitable k -point Gauss–Legendre’s quadrature formula as well as $r = k$ in (24) yields a symplectic method of arbitrary order. This manipulation is very simple and convenient and it opens up the possibility of using high-order symplectic methods to solve the second-order ODEs (5).*

4.3 Convergence analysis of the iteration

It can be observed that usually the TFC method (23) constitutes of a system of implicit equations for the determination of v_i and it requires iterative computation. In this paper, we use the fixed-point iteration in practical computation. The maximum norm for a matrix or a vector is denoted by $\|\cdot\|$. Insofar as the convergence of the fixed-point iteration for the TFC method (23) is concerned, we have the following result.

Theorem 47 *Assume that M is symmetric and positive semi-definite and that f satisfies a Lipschitz condition in the variable q , i.e., there exists a constant L with the property that $\|f(q_1) - f(q_2)\| \leq L \|q_1 - q_2\|$. If*

$$0 < h < \frac{1}{\sqrt{Lr^2 \max_{i=1,\dots,k} c_i^2 \max_{j=1,\dots,k} |b_j|}}, \quad (31)$$

then the fixed-point iteration for the TFC method (23) is convergent. For a quadrature formula, generally speaking, not all of the node points c_i ($i = 1, 2, \dots, k$) are equal to zero and this ensures that $\max_{i=1,\dots,k} c_i^2 \max_{j=1,\dots,k} |b_j| \neq 0$.

Proof. Following Definition 31, the first formula of (23) can be rewritten as

$$Q = \phi_0(c^2V)q_0 + c\phi_1(c^2V)hp_0 + h^2A(V)f(Q), \quad (32)$$

where $c = (c_1, c_2, \dots, c_k)^T$, $Q = (v_1, v_2, \dots, v_k)^T$, $A(V) = (a_{ij}(V))_{k \times k}$ and $a_{ij}(V)$ are defined as

$$a_{ij}(V) := c_i^2 b_j \sum_{l=0}^{r-1} I_{1,l,c_i}(V) \widehat{P}_l(c_j).$$

From (13), it follows that $|\widehat{P}_j| \leq \sqrt{2j+1}$. We then obtain

$$\begin{aligned} \|a_{ij}(V)\| &\leq c_i^2 |b_j| \sum_{l=0}^{r-1} \sqrt{2l+1} \int_0^1 |\widehat{P}_l(c_i z)| \|(1-z)\phi_1((1-z)^2 c_i^2 V)\| dz \\ &\leq c_i^2 |b_j| \sum_{l=0}^{r-1} (2l+1) \int_0^1 \|(1-z)\phi_1((1-z)^2 c_i^2 V)\| dz. \end{aligned}$$

By Proposition 2.1 in [38], we know that $\|\phi_1((1-z)^2 c_i^2 V)\| \leq 1$ and then we get

$$\|a_{ij}(V)\| \leq c_i^2 |b_j| \sum_{l=0}^{r-1} (2l+1) = r^2 c_i^2 |b_j|,$$

which yields $\|A(V)\| \leq r^2 \max_{i=1, \dots, k} c_i^2 \max_{j=1, \dots, k} |b_j|$. Let

$$\varphi(x) = \phi_0(c^2V)q_0 + c\phi_1(c^2V)hp_0 + h^2A(V)f(x).$$

Then

$$\begin{aligned} \|\varphi(x) - \varphi(y)\| &= \|h^2A(V)f(x) - h^2A(V)f(y)\| \leq h^2L \|A(V)\| \|x - y\| \\ &\leq h^2Lr^2 \max_{i=1, \dots, k} c_i^2 \max_{j=1, \dots, k} |b_j| \|x - y\|, \end{aligned}$$

which means that $\varphi(x)$ is a contraction from the assumption (31). The well-known Contraction Mapping Theorem then ensures the convergence of the fixed-point iteration. \square

If the matrix M is not symmetric, it can be observed that the restriction on h becomes $0 < h^2 \|A(V)\| < \frac{1}{L}$. For the RKN-type Fourier collocation method (24), the restriction on h is $0 < h < \frac{1}{\sqrt{L \max_{1 \leq i, j \leq k} \{a_{ij}\}}}$. If the assumption on f in

Theorem 47 is only satisfied in a neighbourhood of the initial value then further restrictions on h are required in order that the argument of f remains in this neighbourhood.

4.4 Stability

In this part we are concerned with the stability. As stated in [15], the linear stability of RKN methods is generally analysed by the test equation

$$y''(t) = -\lambda^2 y(t) \quad \text{with } \lambda > 0. \quad (33)$$

Applying the RKN-type Fourier collocation method (24) to (33) yields the recursion

$$\begin{pmatrix} v_1 \\ hu_1 \end{pmatrix} = Q(\vartheta) \begin{pmatrix} q_0 \\ hp_0 \end{pmatrix},$$

where

$$Q(\vartheta) = \begin{pmatrix} 1 - \vartheta^2 \bar{b}^T N^{-1} e & 1 - \vartheta^2 \bar{b}^T N^{-1} c \\ -\vartheta^2 b^T N^{-1} e & 1 - \vartheta^2 b^T N^{-1} c \end{pmatrix}$$

with $N = I + \vartheta^2 \bar{A}$, $\vartheta = \lambda h$, $e = (1, \dots, 1)^T$.

For an RKN method, we have the following definitions:

- $I_s = \{\vartheta > 0 \mid \rho(Q) < 1\}$ is called the *interval of stability*.
- $I_p = \{\vartheta > 0 \mid \rho(Q) = 1 \text{ and } \text{tr}(Q)^2 < 4 \det(Q)\}$ is called the *interval of periodicity*.

For the TFC method (23), following [50], we use the scalar revised test equation:

$$q''(t) + \omega^2 q(t) = -\epsilon q(t) \quad \text{with } \omega^2 + \epsilon > 0, \quad (34)$$

where ω represents an estimation of the dominant frequency λ and $\epsilon = \lambda^2 - \omega^2$ is the error of that estimation. Applying (23) to (34) produces

$$\begin{aligned} Q &= \phi_0(c^2 V) q_0 + c \phi_1(c^2 V) h p_0 - z A(V) Q, \quad z = \epsilon h^2, \quad V = h^2 \omega^2, \\ v_1 &= \phi_0(V) q_0 + \phi_1(V) h p_0 - z \bar{b}^T(V) Q, \\ hu_1 &= -V \phi_1(V) q_0 + \phi_0(V) h p_0 - z b^T(V) Q, \end{aligned} \quad (35)$$

where c , Q , $A(V)$ are defined in Subsection 4.3 and

$$\begin{aligned} \bar{b}(V) &= \left(b_1 \sum_{j=0}^{r-1} I_{1,j}(V) \hat{P}_j(c_1), \dots, b_k \sum_{j=0}^{r-1} I_{1,j}(V) \hat{P}_j(c_k) \right)^T, \\ b(V) &= \left(b_1 \sum_{j=0}^{r-1} I_{2,j}(V) \hat{P}_j(c_1), \dots, b_k \sum_{j=0}^{r-1} I_{2,j}(V) \hat{P}_j(c_k) \right)^T. \end{aligned}$$

From (35), it follows that

$$\begin{pmatrix} v_1 \\ hu_1 \end{pmatrix} = S(V, z) \begin{pmatrix} q_0 \\ hp_0 \end{pmatrix},$$

where the stability matrix $S(V, z)$ is given by

$$S(V, z) = \begin{pmatrix} \phi_0(V) - z \bar{b}^T(V) N^{-1} \phi_0(c^2 V) & \phi_1(V) - z \bar{b}^T(V) N^{-1} (c \cdot \phi_1(c^2 V)) \\ -V \phi_1(V) - z b^T(V) N^{-1} \phi_0(c^2 V) & \phi_0(V) - z b^T(V) N^{-1} (c \cdot \phi_1(c^2 V)) \end{pmatrix}$$

with $N = I + zA(V)$.

Accordingly, we have the following definitions of stability for our method (23).

Definition 48 $R_s = \{(V, z) \mid V > 0 \text{ and } \rho(S) < 1\}$ is called the stability region of the method (23). $R_p = \{(V, z) \mid V > 0, \rho(S) = 1 \text{ and } \operatorname{tr}(S)^2 < 4 \det(S)\}$ is called the periodicity region of the method (23).

Remark 49 In the previous analysis, we choose the shifted Legendre polynomials as an example of orthonormal basis. It is noted that a different choice of the orthonormal basis can be made and then the above analysis is accordingly modified. We do not discuss this point further in this paper and different choices of the basis will be considered in future investigations.

Remark 410 In this paper, we do not discuss the non-autonomous problem $q''(t) + Mq(t) = f(t, q(t))$ because by appending the equation $t'' = 0$, it can be turned into the autonomous form

$$u''(t) + \widetilde{M}u(t) = g(u(t)).$$

where $u(t) = (t, q^T(t))^T$, $g(u(t)) = (t, f^T(t, q(t)))^T$ and $\widetilde{M} = \begin{pmatrix} 1 & 0_{1 \times d} \\ 0_{d \times 1} & M \end{pmatrix}$. Therefore, our TFC methods and the whole analysis presented in the paper are applicable to the non-autonomous problem $q''(t) + Mq(t) = f(t, q(t))$.

5 Numerical experiments

As an example of the TFC methods, we choose a Gauss–Legendre’s quadrature that is exact for all polynomials of degree ≤ 5 as the quadrature formula in (23), which means that

$$\begin{aligned} c_1 &= \frac{5 - \sqrt{15}}{10}, & c_2 &= \frac{1}{2}, & c_3 &= \frac{5 + \sqrt{15}}{10}, \\ b_1 &= \frac{5}{18}, & b_2 &= \frac{4}{9}, & b_3 &= \frac{5}{18}. \end{aligned} \quad (36)$$

Then we choose $r = 3$ in (23) and denote the corresponding trigonometric Fourier collocation methods as TFC1. Various examples of TFC methods can be obtained by choosing different values of k and r and we do not go further on this point in the paper for brevity. According to the analysis given in Section 4, we have the following result for the method TFC1.

Theorem 51 Let $t_{\text{end}} = Nh$ and apply the method TFC1 with the step size h to (1) in the interval $[0, t_{\text{end}}]$. We denote the numerical solution at Nh by $\omega_N(h)$ and then we have

$$\begin{aligned} \omega_N(h) - \mathbf{y}(Nh) &= \mathcal{O}(h^6), \\ Q(\omega_N(h)) - Q(\mathbf{y}_0) &= \mathcal{O}(h^6), \\ H(\omega_N(h)) - H(\mathbf{y}_0) &= \mathcal{O}(h^6). \end{aligned}$$

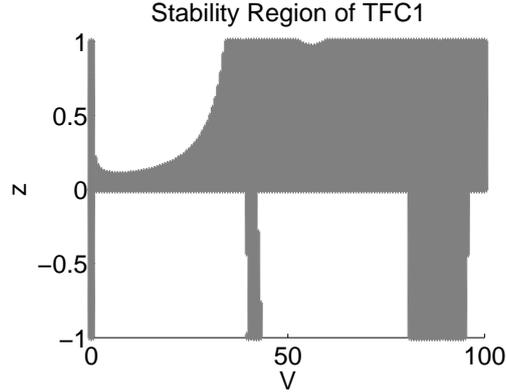


Fig. 1. Stability region (shaded region) of the method TFC1.

Here all the constants implicit in \mathcal{O} are independent of N and h but depend on t_{end} . When $M \rightarrow 0$, the method TFC1 is symplectic and it exactly preserves the quadratic invariant of (5).

For the method TFC1, the stability region is shown in Fig. 1. When $M \rightarrow 0$, the interval of periodicity of RNK-type TFC1 is

$$\left(0, \sqrt{54 - 2\sqrt{489}}\right] \cup \left[\sqrt{10}, 4\sqrt{15/7}\right] \cup \left[2\sqrt{15}, \sqrt{54 + 2\sqrt{489}}\right).$$

In order to show the efficiency and robustness of the method TFC1, the integrators we select for comparison are:

- TFC1: the TFC method derived in this section;
- AAVF–GL: a high precision energy-preserving integrator AAVF–GL using the Gauss–Legendre’s rule (36) in [53];
- LIIIA: the Labatto IIIA method of order six in [22];
- HBVM(3,3): the Hamiltonian Boundary Value Method of order six in [3] which coincides with the three-stage Gauss-Legendre collocation method in [22].

It is noted that these four methods are all implicit and we use a fixed-point iteration in the practical computations. In all the problems, we set 10^{-16} as the error tolerance and 10 as the maximum number of each iteration for showing the efficiency curve as well as energy conservation for a Hamiltonian system. For each problem we also present the requisite total numbers of iterations for each method when choosing different error tolerances in the fixed-point iteration.

Problem 1. Consider the oscillatory nonlinear system (see [16])

$$q'' + \begin{pmatrix} 13 & -12 \\ -12 & 13 \end{pmatrix} q = -\frac{\partial U}{\partial q}$$

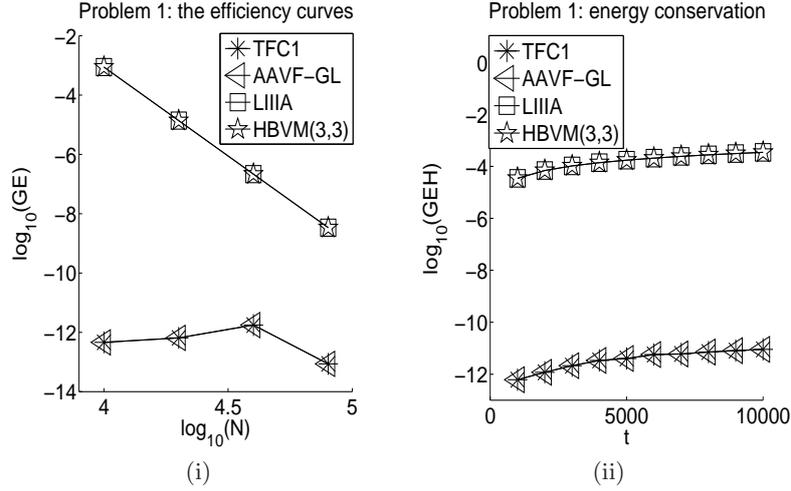


Fig. 2. Results for Problem 1 with the initial condition (37). (i): The logarithm of the global error (GE) over the integration interval against the logarithm of $N = t_{\text{end}}/h$. (ii): The logarithm of the global error of Hamiltonian $GEH = |H_n - H_0|$ against t .

with $U(q) = q_1 q_2 (q_1 + q_2)^3$. Following [16], the initial conditions are chosen as

$$q(0) = \begin{pmatrix} -1 \\ 1 \end{pmatrix}, \quad q'(0) = \begin{pmatrix} -5 \\ 5 \end{pmatrix} \quad (37)$$

such that the analytic solution of this system is $q = \begin{pmatrix} -\cos(5t) - \sin(5t) \\ \cos(5t) + \sin(5t) \end{pmatrix}$. The Hamiltonian is

$$H(q, q') = \frac{1}{2} q'^T q' + \frac{1}{2} q^T \begin{pmatrix} 13 & -12 \\ -12 & 13 \end{pmatrix} q + U(q).$$

First we solve the problem in the interval $[0, 1000]$ with different stepsizes $h = 0.1/2^i$, $i = 0, 1, 2, 3$. The global errors ($GE = \|q(t_{\text{end}}) - v(t_{\text{end}})\|$) are presented in Fig. 2 (i). Then we integrate this problem with the stepsize $h = 0.1$ in the interval $[0, 10000]$. See Fig. 2 (ii) for the energy conservation of different methods. We also solve the problem in the interval $[0, 10]$ with the stepsize $h = 0.01$ and display the total numbers of iterations in Table 1 for different error tolerances (tol) chosen in the fixed-point iteration.

Then we change the initial conditions into

$$q(0) = \begin{pmatrix} -1 \\ 1.1 \end{pmatrix}, \quad q'(0) = \begin{pmatrix} -5 \\ 5 \end{pmatrix} \quad (38)$$

and solve the problem in $[0, 1000]$ with $h = 0.1/2^i$, $i = 0, 1, 2, 3$. See Fig. 3 (i) for the global errors. We also integrate this problem with the stepsize $h = 0.05$ in

Methods	$tol = 1.0e - 006$	$tol = 1.0e - 008$	$tol = 1.0e - 010$	$tol = 1.0e - 012$
TFC1	1000	1000	1000	1000
AAVF-GL	1000	1000	1000	1000
LIIIA	3793	4787	5722	6603
HBVM(3,3)	3600	4547	5331	6000

Table 1. Results for Problem 1 with the initial condition (37): The total numbers of iterations for different error tolerances (tol).

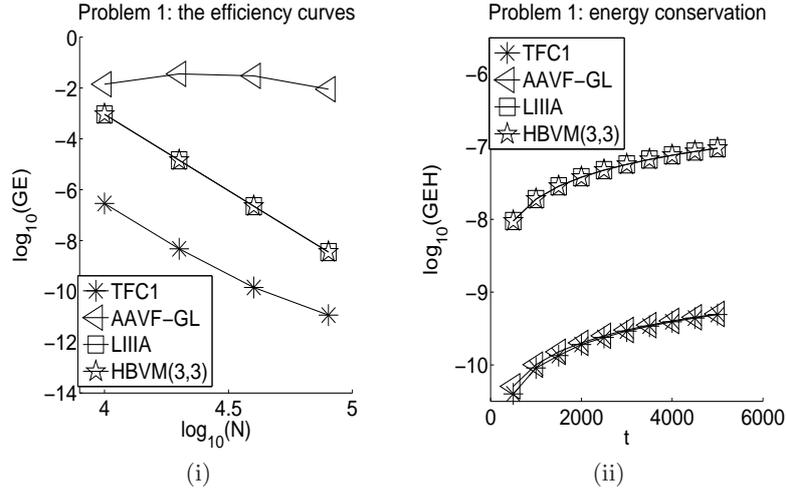


Fig. 3. Results for Problem 1 with the initial condition (38). (i): The logarithm of the global error (GE) over the integration interval against the logarithm of $N = t_{\text{end}}/h$. (ii): The logarithm of the global error of Hamiltonian $GEH = |H_n - H_0|$ against t .

$[0, 5000]$ and present the energy conservation in Fig. 3 (ii). Besides, the problem with the initial condition (38) is solved in $[0, 10]$ with the stepsize $h = 0.01$ and the total numbers of iterations for different error tolerances are listed in Table 2.

Problem 2. Fermi–Pasta–Ulam problem is an important model for simulations in statistical mechanics which is considered in [12, 21, 22, 53, 58]. It is a Hamiltonian system with the Hamiltonian

$$\begin{aligned}
 H(x, y) = & \frac{1}{2} \sum_{i=1}^{2m} y_i^2 + \frac{\omega^2}{2} \sum_{i=1}^m x_{m+i}^2 + \frac{1}{4} \left[(x_1 - x_{m+1})^4 \right. \\
 & \left. + \sum_{i=1}^{m-1} (x_{i+1} - x_{m+i-1} - x_i - x_{m+i})^4 + (x_m + x_{2m})^4 \right].
 \end{aligned}$$

This results in

$$x''(t) + Mx(t) = -\nabla U(x), \quad t \in [0, t_{\text{end}}],$$

Methods	$tol = 1.0e - 006$	$tol = 1.0e - 008$	$tol = 1.0e - 010$	$tol = 1.0e - 012$
TFC1	1000	1516	1964	2000
AAVF-GL	1000	1667	1978	2283
LIIIA	3799	4795	5731	6614
HBVM(3,3)	3612	4560	5356	6000

Table 2. Results for Problem 1 with the initial condition (38): The total numbers of iterations for different error tolerances (tol).

where

$$M = \begin{pmatrix} \mathbf{0}_{m \times m} & \mathbf{0}_{m \times m} \\ \mathbf{0}_{m \times m} & \omega^2 I_{m \times m} \end{pmatrix},$$

$$U(x) = \frac{1}{4} \left[(x_1 - x_{m+1})^4 + \sum_{i=1}^{m-1} (x_{i+1} - x_{m+i-1} - x_i - x_{m+i})^4 + (x_m + x_{2m})^4 \right].$$

We choose

$$m = 3, \omega = 50, x_1(0) = 1, y_1(0) = 1, x_4(0) = \frac{1}{\omega}, y_4(0) = 1,$$

and choose zero for the remaining initial values. The system is integrated in the interval $[0, 100]$ with the stepsizes $h = 0.1/2^k$, $k = 0, 1, 2, 3$. We plot the logarithm of the global errors in Fig. 4 (i). Here it is noted that for LIIIA and HBVM(3,3), the global errors are too large for some values of h , thus we do not plot the corresponding points in Fig. 4 (i). Then we solve this problem in the interval $[0, 10000]$ with the stepsize $h = 0.005$ and present the energy conservation in Fig. 4 (ii). Besides, we solve the problem in the interval $[0, 10]$ with the stepsize $h = 0.01$ to show the convergence rate of iterations for different methods. Table 3 lists the total numbers of iterations for different error tolerances.

Methods	$tol = 1.0e - 006$	$tol = 1.0e - 008$	$tol = 1.0e - 010$	$tol = 1.0e - 012$
TFC1	1164	2000	2036	2992
AAVF-GL	1631	2000	2706	3000
LIIIA	6726	8724	10878	12913
HBVM(3,3)	6353	8529	10789	12821

Table 3. Results for Problem 2: The total numbers of iterations for different error tolerances (tol).

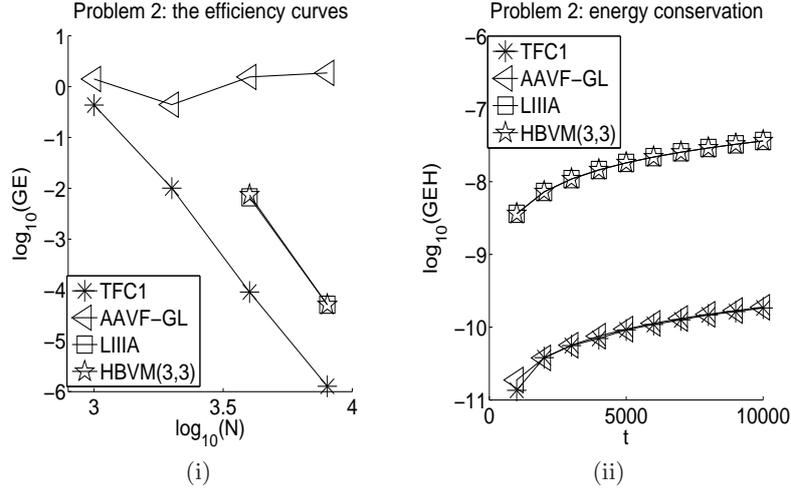


Fig. 4. Results for Problem 2. (i): The logarithm of the global error (GE) over the integration interval against the logarithm of $N = t_{\text{end}}/h$. (ii): The logarithm of the global error of Hamiltonian $GEH = |H_n - H_0|$ against t .

Problem 3. Consider a nonlinear wave equation (see [50])

$$\begin{aligned} \frac{\partial^2 u}{\partial t^2} - gd(x) \frac{\partial^2 u}{\partial x^2} &= \frac{1}{4} \lambda^2(x, u) u, \quad 0 < x \leq b, \quad t > 0, \\ \frac{\partial u}{\partial x}(t, 0) = \frac{\partial u}{\partial x}(t, b) &= 0, \quad u(0, x) = \sin\left(\frac{\pi x}{b}\right), \quad u_t(0, x) = -\frac{\pi}{b} \sqrt{gd(x)} \cos\left(\frac{\pi x}{b}\right), \end{aligned}$$

where $d(x)$ is the depth function given by $d(x) = d_0 \left[2 + \cos\left(\frac{2\pi x}{b}\right) \right]$, g denotes the acceleration of gravity, and $\lambda(x, u)$ is the coefficient of bottom friction defined by $\lambda(x, u) = \frac{g|u|}{C^2 d(x)}$ with Chezy coefficient C .

By using second-order symmetric differences, this problem is converted into a system of ODEs in time

$$\begin{aligned} \frac{d^2 u_i}{dt^2} - gd(x_i) \frac{u_{i+1} - 2u_i + u_{i-1}}{\Delta x^2} &= \frac{1}{4} \lambda^2(x_i, u_i) u_i, \quad 0 < t \leq t_{\text{end}}, \\ u_i(0) = \sin\left(\frac{\pi x_i}{b}\right), \quad u'_i(0) &= -\frac{\pi}{b} \sqrt{gd(x_i)} \cos\left(\frac{\pi x_i}{b}\right), \quad i = 1, 2, \dots, N, \end{aligned}$$

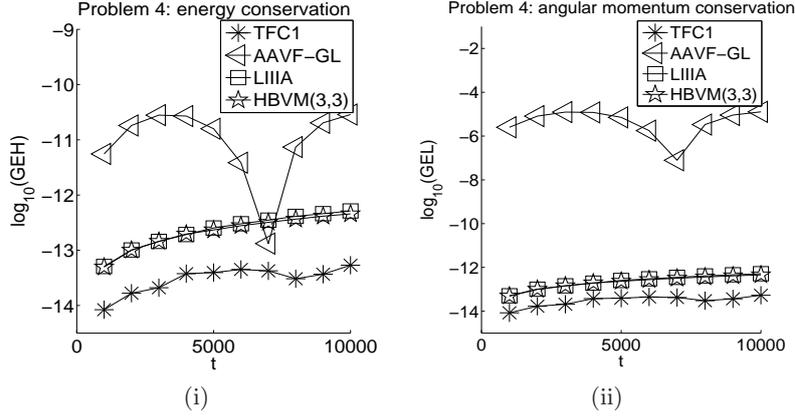


Fig. 6. Results for Problem 4. (i): The logarithm of the global error of Hamiltonian $GEH = |H_n - H_0|$ against t . (ii): The logarithm of the global error of angular momentum $GEL = |L_n - L_0|$ against t .

(i). Table 4 gives the total numbers of iterations when applying the different methods to this problem in the interval $[0, 10]$ with the stepsize $h = 0.01$ and different error tolerances in the fix-point iteration. Since the matrix M in this problem is not symmetric, (39) is not equivalent to a Hamiltonian system and thus there is no Hamiltonian energy to conserve. However, this problem can still be used to compare the accuracy of the numerical solution produced by each method. It also provides an example for supporting the assertion that our TFC methods are still applicable to (1) with a nonsymmetric matrix M . This is the reason why we apply these numerical methods to this non-Hamiltonian problem.

Methods	$tol = 1.0e - 006$	$tol = 1.0e - 008$	$tol = 1.0e - 010$	$tol = 1.0e - 012$
TFC1	1000	1000	1000	1000
AAVF-GL	1000	1000	1000	1000
LIIIA	3000	4020	5896	6989
HBVM(3,3)	3000	4000	5000	6217

Table 4. Results for Problem 3: The total numbers of iterations for different error tolerances (tol).

Now we use the next problem to show that our TFC methods exhibit good performance for a special and important class of second-order ODEs $q'' = f(q)$ ($M = 0$ in (1)).

Problem 4. A perturbed Kepler's problem is given by

$$\begin{aligned} q_1'' &= -\frac{q_1}{(q_1^2 + q_2^2)^{3/2}} - \frac{(2\epsilon + \epsilon^2)q_1}{(q_1^2 + q_2^2)^{5/2}}, & q_1(0) &= 1, & q_1'(0) &= 0, \\ q_2'' &= -\frac{q_2}{(q_1^2 + q_2^2)^{3/2}} - \frac{(2\epsilon + \epsilon^2)q_2}{(q_1^2 + q_2^2)^{5/2}}, & q_2(0) &= 0, & q_2'(0) &= 1 + \epsilon. \end{aligned}$$

The exact solution is $q_1(t) = \cos(t + \epsilon t)$, $q_2(t) = \sin(t + \epsilon t)$. The Hamiltonian is

$$H = \frac{1}{2}(q_1'^2 + q_2'^2) - \frac{1}{\sqrt{q_1^2 + q_2^2}} - \frac{(2\epsilon + \epsilon^2)}{3(q_1^2 + q_2^2)^{3/2}}.$$

The system also has the angular momentum $L = q_1 q_2' - q_2 q_1'$ as a first integral. We take the parameter value $\epsilon = 10^{-3}$.

First we solve the problem in the interval $[0, 1000]$ with different stepsizes $h = 0.1/2^{i-1}$, $i = 0, 1, 2, 3$ and present the global errors in Fig. 5 (ii). Then we integrate this problem with the stepsize $h = 0.1$ in the interval $[0, 10000]$ and devote Fig. 6 for the energy conservation and angular momentum conservation. This problem is also solved in the interval $[0, 10]$ with the stepsize $h = 0.01$ and Table 5 lists the total numbers of iterations for different error tolerances. It is noted that the RKN-type Fourier collocation methods (24) are closely related to the classical Gauss-Legendre collocation methods adapted to second order problems and their computational complexity is likely the same. However, it follows from the results of this problem that the new RKN-type Fourier collocation methods share more favorable error constants than the Gauss-Legendre collocation methods.

Methods	$tol = 1.0e - 006$	$tol = 1.0e - 008$	$tol = 1.0e - 010$	$tol = 1.0e - 012$
TFC1	1000	2000	2000	2000
AAVF-GL	1000	2000	2000	3000
LIIIA	2000	3000	4000	5000
HBVM(3,3)	2000	3000	4000	5000

Table 5. Results for Problem 4: The total numbers of iterations for different error tolerances (tol).

By the results of the numerical experiments, it can be observed clearly that our method provides a considerably more accurate numerical solution than other methods and preserves well the Hamiltonian energy. Moreover, our method requires less fixed-point iterations than other methods, which is significant in long-term computations.

6 Conclusions

This paper presents a framework for the derivation and analysis of a novel class of TFC methods (23) for the multi-frequency oscillatory system (1) or (2). These new collocation methods are based on the variation-of-constants formula and a local Fourier expansion of the problem. Based on the TFC methods (23), a special RKN-type Fourier collocation method (24) is designed for solving a special and important class of second-order ODEs (5). It has been shown that the new TFC methods can have arbitrary order and we can obtain symplectic RKN-type Fourier collocation methods (24) with arbitrary order of accuracy for (5) in a very convenient and simple way. Numerical experiments carried out in this paper demonstrate clearly that the novel TFC methods have excellent numerical behaviour in comparison with some existing methods in the scientific literature.

It is noted that the approach developed in this paper could be used for differential equations of the more general form $q'(t) + Aq(t) = f(q(t))$ with a matrix A .

Acknowledgments. Bin Wang and Xinyuan Wu was supported in part by the Natural Science Foundation of China under Grants 11271186 and 11401333, by NSFC and RS International Exchanges Project under Grant 113111162, by the Specialized Research Foundation for the Doctoral Program of Higher Education under Grant 20130091110041, by the 985 Project at Nanjing University under Grant 9112020301, by A Project Funded by the Priority Academic Program Development of Jiangsu Higher Education Institutions. Bin Wang is sincerely thankful to Numerical Analysis Group at University of Cambridge since the work was partly done when he was studying in this group as a visiting student. The revised version of the manuscript was completed during Bin Wang and Xinyuan Wu were visiting to Numerical Analysis Group at University of Cambridge in July and August, 2014.

The authors are sincerely thankful to two anonymous reviewers for their valuable suggestions, which help improve the presentation of the manuscript significantly.

Appendix A. Proof of Theorem 43

By virtue of Lemma 41, (27) and (28), one has

$$\begin{aligned}
H(\omega(h)) - H(\mathbf{y}_0) &= h \int_0^1 \nabla H(\omega(\xi h))^T \omega'(\xi h) d\xi \\
&= h \int_0^1 \left((Mv(\xi h) - \sum_{j=0}^{\infty} \hat{P}_j(\xi) \gamma_j(v))^T, u(\xi h)^T \right) \\
&\quad \cdot \left(-Mv(\xi h) + \sum_{j=0}^{r-1} \hat{P}_j(\xi) \sum_{l=1}^k b_l \hat{P}_j(c_l) f(v(c_l h)) \right) d\xi \\
&= h \int_0^1 u(\xi h)^T \left(\sum_{j=0}^{r-1} \hat{P}_j(\xi) \sum_{l=1}^k b_l \hat{P}_j(c_l) f(v(c_l h)) - \sum_{j=0}^{\infty} \hat{P}_j(\xi) \gamma_j(v) \right) d\xi \\
&= h \int_0^1 u(\xi h)^T \left(- \sum_{j=0}^{r-1} \hat{P}_j(\xi) \Delta_j(h) - \sum_{j=r}^{\infty} \hat{P}_j(\xi) \gamma_j(v) \right) d\xi \\
&= -h \sum_{j=0}^{r-1} \int_0^1 u(\xi h)^T \hat{P}_j(\xi) d\xi \Delta_j(h) - h \sum_{j=r}^{\infty} \int_0^1 u(\xi h)^T \hat{P}_j(\xi) d\xi \gamma_j(v) \\
&= h \sum_{j=0}^{r-1} \mathcal{O}(h^j \times h^{m-j}) + h \sum_{j=r}^{\infty} \mathcal{O}(h^j \times h^j) = \mathcal{O}(h^{m+1}) + \mathcal{O}(h^{2r+1}).
\end{aligned}$$

□

Appendix B. Proof of Theorem 44

From $Q(\mathbf{y}) = q^T Dp$ and $D^T = -D$, it follows that

$$\begin{aligned}
Q(\omega(h)) - Q(\mathbf{y}_0) &= h \int_0^1 \nabla Q(\omega(\xi h))^T \omega'(\xi h) d\xi \\
&= h \int_0^1 \left(-u(\xi h)^T D, v(\xi h)^T D \right) \left(-Mv(\xi h) + \sum_{j=0}^{r-1} \hat{P}_j(\xi) \sum_{l=1}^k b_l \hat{P}_j(c_l) f(v(c_l h)) \right) d\xi.
\end{aligned}$$

Since $q^T D(f(q) - Mq) = 0$ for any $q \in \mathbb{R}^d$, we obtain

$$\begin{aligned}
 Q(\omega(h)) - Q(\mathbf{y}_0) &= h \int_0^1 v(\xi h)^T D \left(-Mv(\xi h) + \sum_{j=0}^{r-1} \widehat{P}_j(\xi) \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v(c_l h)) \right) d\xi \\
 &= h \int_0^1 v(\xi h)^T D \left(\sum_{j=0}^{r-1} \widehat{P}_j(\xi) \sum_{l=1}^k b_l \widehat{P}_j(c_l) f(v(c_l h)) - \sum_{j=0}^{\infty} \widehat{P}_j(\xi) \gamma_j(v) \right) d\xi \\
 &= h \int_0^1 v(\xi h)^T D \left(-\sum_{j=0}^{r-1} \widehat{P}_j(\xi) \Delta_j(h) - \sum_{j=r}^{\infty} \widehat{P}_j(\xi) \gamma_j(v) \right) d\xi \\
 &= -h \sum_{j=0}^{r-1} \int_0^1 v(\xi h)^T D \widehat{P}_j(\xi) d\xi \Delta_j(h) - h \sum_{j=r}^{\infty} \int_0^1 v(\xi h)^T D \widehat{P}_j(\xi) d\xi \gamma_j(v) \\
 &= h \sum_{j=0}^{r-1} \mathcal{O}(h^j \times h^{m-j}) + h \sum_{j=r}^{\infty} \mathcal{O}(h^j \times h^j) = \mathcal{O}(h^{m+1}) + \mathcal{O}(h^{2r+1}).
 \end{aligned}$$

□

References

1. L. Brugnano and F. Iavernaro. Line integral methods which preserve all invariants of conservative problems. *J. Comput. Appl. Math.* 236 (2012), 3905–3919.
2. L. Brugnano, F. Iavernaro and D. Trigiante. A note on the efficient implementation of Hamiltonian BVMs. *J. Comput. Appl. Math.* 236 (2011), 375–383.
3. L. Brugnano, F. Iavernaro and D. Trigiante. A simple framework for the derivation and analysis of effective one-step methods for ODEs. *Appl. Math. Comput.* 218 (2012), 8475–8485.
4. L. Brugnano, F. Iavernaro and D. Trigiante. Energy and Quadratic Invariants Preserving integrators based upon Gauss collocation formulae. *SIAM J. Numer. Anal.* 50 (2012), 2897–2916.
5. R.W. Butler and A.T.A. Wood. Laplace approximations for hypergeometric functions with matrix argument. *Ann. Statist.* 30 (2002), 1155–1177.
6. E. Celledoni, R.I. McLachlan, B. Owren and G.R.W. Quispel. Energy-Preserving Integrators and the Structure of B-series. *Found. Comput. Math.* 10 (2010), 673–693.
7. P. Chartier and A. Murua. Preserving first integrals and volume forms of additively split systems. *IMA J. Numer. Anal.* 27 (2007), 381–405.
8. J.L. Cieslinski and B. Ratkiewicz. Energy-preserving numerical schemes of high accuracy for one-dimensional Hamiltonian systems. *J. Phys. A: Math. Theor.* 44 (2011), 155206.
9. D. Cohen. Conservation properties of numerical integrators for highly oscillatory Hamiltonian systems. *IMA J. Numer. Anal.* 26 (2006), 34–59.
10. D. Cohen and E. Hairer. Linear energy-preserving integrators for Poisson systems. *BIT* 51 (2011), 91–101.
11. D. Cohen, E. Hairer and C. Lubich. Numerical energy conservation for multi-frequency oscillatory differential equations. *BIT* 45 (2005), 287–305.

12. D. Cohen, T. Jahnke, K. Lorenz and C. Lubich. Numerical integrators for highly oscillatory Hamiltonian systems: a review, in *Analysis, Modeling and Simulation of Multiscale Problems* (A. Mielke, ed.). Springer, Berlin (2006), 553–576 .
13. M. Dahlby, B. Owren and T. Yaguchi. Preserving multiple first integrals by discrete gradients. *J. Phys. A: Math. Theor.* 44 (2011), 305205.
14. J.M. Franco. Runge-Kutta-Nyström methods adapted to the numerical integration of perturbed oscillators. *Comput. Phys. Comm.* 147 (2002), 770–787.
15. J.M. Franco. Stability of explicit ARKN methods for perturbed oscillators. *J. Comput. Appl. Math.* 173 (2005), 389–396.
16. J.M. Franco. New methods for oscillatory systems based on ARKN methods. *Appl. Numer. Math.* 56 (2006), 1040–1053.
17. A. García, P. Martín and A.B. González. New methods for oscillatory problems based on classical codes. *Appl. Numer. Math.* 42 (2002), 141–157.
18. B. García-Archilla, J.M. Sanz-Serna and R.D. Skeel. Long-time-step methods for oscillatory differential equations. *SIAM J. Sci. Comput.* 20 (1999), 930–963.
19. R. Gutiérrez, J. Rodríguez and A.J. Sáez. Approximation of hypergeometric functions with matricial argument through their development in series of zonal polynomials. *Electron. Trans. Numer. Anal.* 11 (2000), 121–130.
20. E. Hairer. Energy-preserving variant of collocation methods. *JNAIAM J. Numer. Anal. Ind. Appl. Math.* 5 (2010), 73–84.
21. E. Hairer and C. Lubich. Long-time energy conservation of numerical methods for oscillatory differential equations. *SIAM J. Numer. Anal.* 38 (2000), 414–441.
22. E. Hairer, C. Lubich and G. Wanner. *Geometric Numerical Integration: Structure-Preserving Algorithms for Ordinary Differential Equations*. 2nd edn. Springer-Verlag, Berlin, Heidelberg, 2006.
23. E. Hairer, R.I. McLachlan and R.D. Skeel. On energy conservation of the simplified Takahashi-Imada method. *Math. Model. Numer. Anal.* 43 (2009), 631–644.
24. E. Hairer, S.P. Nørsett and G. Wanner. *Solving Ordinary Differential Equations I: Nonstiff Problems*. Springer-Verlag, Berlin, 1993.
25. J.K. Hale. *Ordinary Differential Equations*. Roberte E. Krieger Publishing company, Huntington, New York, 1980.
26. M. Hochbruck and C. Lubich. A Gautschi-type method for oscillatory second-order differential equations. *Numer. Math.* 83 (1999), 403–426.
27. M. Hochbruck and A. Ostermann. Explicit exponential Runge–Kutta methods for semilinear parabolic problems. *SIAM J. Numer. Anal.* 43 (2005), 1069–1090.
28. M. Hochbruck and A. Ostermann. Exponential integrators. *Acta Numer.* 19 (2010), 209–286.
29. M. Hochbruck, A. Ostermann and J. Schweitzer. Exponential rosenbrock-type methods. *SIAM J. Numer. Anal.* 47 (2009), 786–803.
30. F. Iavernaro and B. Pace. Conservative Block-Boundary Value Methods for the solution of Polynomial Hamiltonian Systems. *AIP Conf. Proc.* 1048 (2008), 888–891.
31. F. Iavernaro and D. Trigiante. High-order symmetric schemes for the energy conservation of polynomial Hamiltonian problems. *JNAIAM J. Numer. Anal. Ind. Appl. Math.* 4 (2009), 787–101.
32. A. Iserles. *A First Course in the Numerical Analysis of Differential Equations*. 2nd edn. Cambridge University Press, Cambridge, 2008.
33. A. Iserles, G.R.W. Quispel and P.S.P. Tse. B-series methods cannot be volume-preserving. *BIT* 47 (2007), 351–378.
34. A. Iserles and A. Zanna. Preserving algebraic invariants with Runge-Kutta methods. *J. Comput. Appl. Math.* 125 (2000), 69–81.

35. P. Koev and A. Edelman. The efficient evaluation of the hypergeometric function of a matrix argument. *Math. Comput.* 75 (2006), 833–846.
36. M. Leok and T. Shingel. Prolongation-collocation variational integrators. *IMA J. Numer. Anal.* 32 (2012), 1194–1216.
37. J. Li, B. Wang, X. You and X. Wu. Two-step extended RKN methods for oscillatory systems. *Comput. Phys. Comm.* 182 (2011), 2486–2507.
38. J. Li and X. Wu. Adapted Falkner-type methods solving oscillatory second-order differential equations. *Numer. Algo.* 62 (2013), 355–381.
39. R.I. McLachlan, G.R.W. Quispel and P.S.P. Tse. Linearization-preserving self-adjoint and symplectic integrators. *BIT* 49 (2009), 177–197.
40. M. Petkovšek, H.S. Wilf and D. Zeilberger. *A=B*, AK Peters Ltd. Wellesley, MA, 1996.
41. G.R.W. Quispel and D.I. McLaren. A new class of energy-preserving numerical integration methods. *J. Phys. A* 41 (2008), 045206.
42. E.D. Rainville. *Special functions*. Macmillan, New York, 1960.
43. S. Reich. Symplectic integration of constrained Hamiltonian systems by composition methods. *SIAM J. Numer. Anal.* 33 (1996), 475–491.
44. S. Reich. On higher-order semi-explicit symplectic partitioned Runge–Kutta methods for constrained Hamiltonian systems. *Numer. Math.* 76 (1997), 231–247.
45. D.S.P. Richards. High-Dimensional Random Matrices from the Classical Matrix Groups, and Generalized Hypergeometric Functions of Matrix Argument. *Symmetry* 3 (2011), 600–610.
46. J.M. Sanz-Serna. Symplectic integrators for Hamiltonian problems: an overview. *Acta Numer.* 1 (1992), 243–286.
47. C.W. Scherr and E.V. Ivash. Associated Legendre Functions. *Am. J. Phys.* 31 (1963), 753.
48. L.J. Slater. *Generalized hypergeometric functions*. Cambridge University Press, Cambridge, 1966.
49. Sun Geng. Construction of high order symplectic Runge–Kutta methods. *J. Comput. Math.* 11 (1993), 250–260.
50. P.J. Van der Houwen and B.P. Sommeijer. Explicit Runge–Kutta (-Nyström) methods with reduced phase errors for computing oscillating solutions. *SIAM J. Numer. Anal.* 24 (1987), 595–617.
51. H. Van de Vyver. A fourth-order symplectic exponentially fitted integrator. *Comput. Phys. Comm.* 174 (2006), 115–130.
52. B. Wang, K. Liu and X. Wu. A Filon-type asymptotic approach to solving highly oscillatory second-order initial value problems. *J. Comput. Phys.* 243 (2013), 210–223.
53. B. Wang and X. Wu. A new high precision energy-preserving integrator for system of oscillatory second-order differential equations. *Phys. Lett. A* 376 (2012), 1185–1190.
54. B. Wang, X. Wu and H. Zhao. Novel improved multidimensional Strömer-Verlet formulas with applications to four aspects in scientific computation. *Math. Comput. Modell.* 57 (2013), 857–872.
55. K. Wright. Some relationships between implicit Runge-Kutta, collocation and Lanczos τ methods, and their stability properties. *BIT* 10 (1970), 217–227.
56. X. Wu and B. Wang. Multidimensional adapted Runge-Kutta-Nyström methods for oscillatory systems. *Comput. Phys. Comm.* 181 (2010), 1955–1962.
57. X. Wu, B. Wang and W. Shi. Efficient energy-preserving integrators for oscillatory Hamiltonian systems. *J. Comput. Phys.* 235 (2013), 587–605.

58. X. Wu, B. Wang and J. Xia. Explicit symplectic multidimensional exponential fitting modified Runge-Kutta-Nyström methods. *BIT* 52 (2012), 773–795.
59. X. Wu, X. You, W. Shi and B. Wang. ERKN integrators for systems of oscillatory second-order differential equations. *Comput. Phys. Comm.* 181 (2010), 1873–1887.
60. X. Wu, X. You and B. Wang. *Structure-Preserving Algorithms for Oscillatory Differential Equations*. Springer-Verlag, Berlin, Heidelberg, 2013.
61. X. Wu, X. You and J. Xia. Order conditions for ARKN methods solving oscillatory systems. *Comput. Phys. Comm.* 180 (2009), 2250–2257.