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Citation: The Journal of Chemical Physics **146**, 114109 (2017); doi: 10.1063/1.4978410 View online: http://dx.doi.org/10.1063/1.4978410 View Table of Contents: http://aip.scitation.org/toc/jcp/146/11 Published by the American Institute of Physics

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Symplectic time-average propagators for the Schrödinger equation with a time-dependent Hamiltonian

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(Received 28 December 2016; accepted 27 February 2017; published online 21 March 2017)

Several symplectic splitting methods of orders four and six are presented for the step-by-step time numerical integration of the Schrödinger equation when the Hamiltonian is a general explicitly time-dependent real operator. They involve linear combinations of the Hamiltonian evaluated at some intermediate points. We provide the algorithm and the coefficients of the methods, as well as some numerical examples showing their superior performance with respect to other available schemes. *Published by AIP Publishing*. [http://dx.doi.org/10.1063/1.4978410]

I. INTRODUCTION

Many quantum mechanical problems of physical interest involve time-dependent potentials. One has to deal then with the time-dependent Schrödinger equation ($\hbar = 1$)

$$i\frac{\partial}{\partial t}\psi(x,t) = \hat{H}(t)\psi(x,t), \qquad \psi(x,0) = \psi_0(x), \qquad (1)$$

where the Hamiltonian is given by

$$\hat{H}(t) = \hat{T} + \hat{V}(t) = -\Delta/(2\mu) + \hat{V}(t),$$
 (2)

where μ is the reduced mass, Δ is the Laplacian operator, and $\hat{V}(t)$ depends explicitly on time. If the potential changes slowly with time, a usual technique for getting approximate numerical solutions consists of subdividing the time integration interval in a number of sufficiently short subintervals of length τ so that \hat{V} is approximately constant in each subinterval and then solve (1) and (2) with the time-varying potential $\hat{V}(t)$ replaced by

$$\overline{V}(t) = \overline{V}_k \quad \text{for} \quad t \in [(k-1)\tau, k\tau]$$
(3)

with time-independent potentials \overline{V}_k . In this way, one gets $\psi_k \approx \psi(\cdot, t_k)$ at times $t_k = k\tau, k = 1, 2, 3, ...$ and eventually reachess the desired final time. The time-independent potential \overline{V}_k can be chosen in several ways: as the average on the subinterval, $\overline{V}_k = (1/\tau) \int_{t_{k-1}}^{t_k} \hat{V}(t) dt$, as the value at the midpoint, $\overline{V}_k = \hat{V}(t_{k-1} + \tau/2)$, etc.^{1–3}

The procedure typically requires discretizing first the Schrödinger equation (1) in space, for which several techniques are widely used: finite differences, spectral methods based on collocation with trigonometric polynomials, Galerkin method with a Hermite basis, etc, both in one or several dimensions (see Ref. 4 and references therein). Once this process has been carried out, one is led to a linear ordinary differential equation (ODE) of the form

$$i\frac{a}{dt}u(t) = H(t)u(t), \qquad u(0) = u_0 \in \mathbb{C}^N, \tag{4}$$

where H(t) is a Hermitian piecewise constant $N \times N$ matrix,

$$H(t) = H_k \quad \text{for} \quad t \in [(k-1)\tau, k\tau], \tag{5}$$

and each $H_k = T + V_k$ is a discretized counterpart of the operator $\overline{H}_k \equiv \hat{T} + \overline{V}_k$. Its solution $u(t) \in \mathbb{C}^N$ represents a fully discretized version of the wave function $\psi(x, t)$ at *N* space grid points, with *N* usually a large number. The values $u_k = u(k\tau)$ of the solution u(t) of (4) evaluated at the time grid points $t = k \tau$ (k = 1, 2, ...) can be computed as

$$u_k = e^{-\iota \tau H_k} u_{k-1}, (6)$$

but instead of evaluating exactly the matrix exponentials in (6) for determining u_k , two alternatives are widely used: (i) the exponential is approximated by the so-called unitary split operator algorithms, i.e., compositions of the form

$$e^{-i\tau H_k} \approx e^{-ib_m\tau V_k} e^{-ia_m\tau T} \cdots e^{-ib_1\tau V_k} e^{-ia_1\tau T}, \qquad (7)$$

where $\{a_i, b_i\}$ are appropriately chosen real coefficients,^{5–8} and (ii) the action of the exponential on the vector u_{k-1} is approximated as

$$e^{-i\tau H_k}u_{k-1} \approx P_m(\tau H_k)u_{k-1},\tag{8}$$

where $P_m(y)$ is a polynomial of degree *m* in *y* that approximates the exponential e^{-iy} . Standard choices for such $P_m(y)$ are truncated Taylor or Chebyshev series expansion of e^{-iy} for an appropriate real interval of *y* or a Lanczos approximation.⁹

When H_k is an arbitrary real symmetric matrix, the use of symplectic splitting methods has allowed the present authors to develop an algorithm that approximately computes $e^{-i\tau H_k v}$ for given $\tau \in \mathbb{R}$ and $v \in \mathbb{C}^N$ within a prescribed error tolerance with a minimum number of matrix-vector products $H_k v$. The algorithm is more efficient than Chebyshev methods under the same conditions for all tolerances and time integration intervals.¹⁰

If the Hamiltonian does not evolve so slowly with time, the previous procedure is not particularly appropriate,

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however. Notice that either averaging \hat{H} on each subinterval or taking each value at the midpoint produces only a second order of approximation in τ , whereas both the Chebyshev methods and the algorithm based on symplectic splitting methods have been designed to be used with large time steps. In consequence, the performance of such methods for approximately computing $e^{-i\tau H_k v}$ degrades with a reduced τ , whereas large values of τ produce results with low accuracy due to a poor approximation of the original time-dependent potential $\hat{V}(t)$ by the piecewise constant potential (3). In that case, it would be more appropriate to use the numerical integrators designed to solve a linear system of the form (4) with H(t), a continuously time-dependent matrix.

Many algorithms specifically designed for Equation (4) exist indeed in the literature. We mention in particular splitoperator methods,¹¹ Runge–Kutta¹² and symplectic partitioned Runge–Kutta methods,¹³ and a combination of a 4th-order Magnus method with the Lanczos algorithm² and the so-called (t, t') method.¹⁴ In the paper,² an exhaustive comparison is carried out of these schemes. Polynomial approximations to the propagator based on Taylor¹⁵ and Chebyshev polynomials^{16–19} have also been proposed to deal with this problem. The review¹⁹ provides a thorough presentation of one such class of polynomial approximations which are in principle also valid for nonlinear versions of the Schrödinger equation.

Motivated by the excellent performance exhibited by symplectic splitting methods for the computation of $e^{-i\tau H}v$ for real symmetric matrices H, our purpose here is to adapt that methodology to the more general case of a time-dependent matrix H(t) by using some techniques introduced in Refs. 20 and 21. The resulting algorithms only involve matrix-vector products of the form H(t)v at different values of t. Here we mainly focus on the application to linear ordinary differential equation (4) obtained by a space discretization of a Hamiltonian operator $\hat{H}(t)$ in (1) and (2) with a time-dependent potential, although the integration schemes presented in this paper are valid for arbitrary linear ODE systems of the form (4) with a time-dependent real matrix H(t), provided that H(t) is reasonably smooth in t.

II. SYMPLECTIC SPLITTING METHODS

Before presenting the proposed numerical integrators for systems of the form (4), we first recall the relatively simpler case of a constant real matrix $H(t) \equiv \tilde{H}$ and the existing schemes within the same family designed to approximate $u(t) = e^{-it\tilde{H}}u_0$.

A. Time-independent Hamiltonian

The class of symplectic splitting methods developed in Refs. 22 and 23 as an alternative to Chebyshev polynomial approximations of $e^{-i\tau \tilde{H}}$ is obtained by applying special purpose integrators to the equation

$$i\frac{d}{dt}u = \widetilde{H}u, \qquad u(0) = u_0 \tag{9}$$

expressed in terms of the real and imaginary part of the vector $u \in \mathbb{C}^N$, that is,

$$\begin{pmatrix} q \\ p \end{pmatrix} = \begin{pmatrix} \operatorname{Re}(u) \\ \operatorname{Im}(u) \end{pmatrix} \in \mathbb{R}^{2N},$$

so that Eq. (9) is equivalent to

 $\frac{d}{dt}\begin{pmatrix} q\\ p \end{pmatrix} = \begin{pmatrix} 0 & \widetilde{H}\\ 0 & 0 \end{pmatrix} \begin{pmatrix} q\\ p \end{pmatrix} + \begin{pmatrix} 0 & 0\\ -\widetilde{H} & 0 \end{pmatrix} \begin{pmatrix} q\\ p \end{pmatrix}.$ (10)

Application of a splitting scheme with appropriate coefficients

$$a_1, b_1, a_2, \dots, a_m, b_m, a_{m+1} \in \mathbb{R}$$
 (11)

leads (see Refs. 10 and 24-26) to an approximation of the solution operator of (10) of the form

$$\begin{pmatrix} I \ \tau a_{m+1}\widetilde{H} \\ 0 \ I \end{pmatrix} \begin{pmatrix} I \ 0 \\ -\tau b_m \widetilde{H} \ I \end{pmatrix} \cdots \begin{pmatrix} I \ 0 \\ -\tau b_1 \widetilde{H} \ I \end{pmatrix} \begin{pmatrix} I \ \tau a_1 \widetilde{H} \\ 0 \ I \end{pmatrix}.$$
 (12)

If this procedure is applied to get a numerical approximation of u(t) in the time interval $t \in [0, t_f]$ using M steps with time step $\tau = t_f/M$, then the following algorithm results:

$$v = H \operatorname{Im}(u_{0})$$

do $k = 0, M - 1$
 $q_{0} = \operatorname{Re}(u_{k}), \quad p_{0} = \operatorname{Im}(u_{k})$
 $q_{1} = q_{0} + \tau a_{1}v$
do $i = 1, m$
 $p_{i} = p_{i-1} - \tau b_{i}\widetilde{H}q_{i}$ (13)
 $v = \widetilde{H}p_{i}$
 $q_{i+1} = q_{i} + \tau a_{i+1}v$
enddo
 $u_{k} = q_{m+1} + ip_{m}$
enddo.

The coefficients of method (12) usually verify $a_{m+2-j} = a_j$ and $b_{m+1-j} = b_j$. In other words, composition (12) is left-right palindromic so that the scheme preserves time-symmetry. Although it is neither unitary nor unconditionally stable (as is the case of method (7)), it is symplectic and conjugate to a unitary scheme so that neither the average error in energy nor the norm of the approximate solution grows with time. Symplectic integrators preserve most qualitative properties of the exact solution and belong to a class of geometric numerical integrators^{27–29} showing, in general, a more favorable error propagation than standard (e.g., Runge-Kutta, Taylor, multistep, etc.) integrators.

The error analysis carried out in Ref. 26, in particular the sharp error estimates obtained there, provides appropriate criteria to construct splitting methods of the form (12) possessing a large stability interval and especially adapted to different regularity conditions by optimizing the main error terms. This analysis has allowed the present authors to construct a set of symplectic splitting methods designed to be used under different precision requirements and time intervals, as well as a technique to select among them the most efficient scheme to approximate $e^{-i\tau \tilde{H}}v$ with the desired accuracy and a reduced computational cost. The resulting algorithm is presented in Ref. 10, where its performance is also compared with methods based on Chebyshev polynomials.

B. Time-dependent Hamiltonian

In the following, we address the more general problem (4) with a generic time-dependent real matrix H(t) and show how the previous splitting scheme (12) can be generalized to this setting. Here we only provide a brief summary of the main steps involved and refer the reader to Refs. 20 and 21 for a more detailed treatment.

We again express $u \in \mathbb{C}^N$ in terms of the real and imaginary part so that Eq. (4) is equivalently written as

$$\frac{d}{dt}\begin{pmatrix} q\\ p \end{pmatrix} = \begin{pmatrix} 0 & H(t)\\ 0 & 0 \end{pmatrix} \begin{pmatrix} q\\ p \end{pmatrix} + \begin{pmatrix} 0 & 0\\ -H(t) & 0 \end{pmatrix} \begin{pmatrix} q\\ p \end{pmatrix}.$$
 (14)

Let $K(t, \tau)$ denote the solution operator so that any solution (q(t), p(t)) of (14) is formally expressed as

$$\begin{pmatrix} q(t+\tau)\\ p(t+\tau) \end{pmatrix} = K(t,\tau) \begin{pmatrix} q(t)\\ p(t) \end{pmatrix}$$

for all $t, \tau \in \mathbb{R}$.

To determine the solution u(t) of the linear ODE system (14) at times $t = t_k$ (k = 1, 2, 3, ...), we will obtain approximations $q_k \approx q(t_k) = \text{Re}(u(t_k))$, $p_k \approx p(t_k) = \text{Im}(u(t_k))$ for k = 0, 1, 2, ... as

$$\begin{pmatrix} q_{k+1} \\ p_{k+1} \end{pmatrix} = \widetilde{K}(t_k, \tau) \begin{pmatrix} q_k \\ p_k \end{pmatrix}$$

(starting from $q_0 = \operatorname{Re}(u_0)$, $p_0 = \operatorname{Im}(u_0)$), where $\widetilde{K}(t, \tau)$ is some approximation of the solution operator $K(t, \tau)$. More specifically, our proposal is to take $\widetilde{K}(t, \tau)$ as

$$\widetilde{K}(t,\tau) = \begin{pmatrix} I \ \widetilde{H}_{m+1}^{A} \\ 0 \ I \end{pmatrix} \begin{pmatrix} I \ 0 \\ -\widetilde{H}_{m}^{B} \ I \end{pmatrix} \cdots \begin{pmatrix} I \ 0 \\ -\widetilde{H}_{1}^{B} \ I \end{pmatrix} \begin{pmatrix} I \ \widetilde{H}_{1}^{A} \\ 0 \ I \end{pmatrix}, \quad (15)$$

where

$$\widetilde{H}_i^A = \tau \sum_{j=1}^n a_{i,j} H(t+c_j\tau), \quad \widetilde{H}_i^B = \tau \sum_{j=1}^n b_{i,j} H(t+c_j\tau), \quad (16)$$

for some *n* and appropriately chosen coefficients $a_{i,j}, b_{i,j}, c_j \in \mathbb{R}$. Notice that the composition (15) is similar in form to (12), but now the matrix \tilde{H} in each factor in (15) is replaced by a different (weighted) time-average of H(t). The approximation (15) is said to have *m* stages.

When schemes (15) and (16) with n = 3 are considered as the approximate solution operator, then the following algorithm results for determining the approximations $u_k \approx u(t_k)$, k = 1, 2, 3, ...

$$H^{A} = 0, \quad u_{k} = u_{0}$$

do $k = 0, M - 1$
 $t_{k} = t_{0} + k \tau, \quad q_{0} = \operatorname{Re}(u_{k}), \quad p_{0} = \operatorname{Im}(u_{k})$
 $H_{1} = H(t_{k} + c_{1}\tau), \quad H_{2} = H(t_{k} + c_{2}\tau), \quad H_{3} = H(t_{k} + c_{3}\tau)$
 $q_{1} = q_{0} + \tau \left(H^{A} + a_{1,1}H_{1} + a_{1,2}H_{2} + a_{1,3}H_{3}\right)p_{0}$
do $i = 1, m - 1$
 $p_{i} = p_{i-1} - \tau \left(b_{i,1}H_{1} + b_{i,2}H_{2} + b_{i,3}H_{3}\right)q_{i}$
 $q_{i+1} = q_{i} + \tau \left(a_{i+1,1}H_{1} + a_{i+1,2}H_{2} + a_{i+1,3}H_{3}\right)p_{i}$
enddo
 $p_{m} = p_{m-1} - \tau \left(b_{m,1}H_{1} + b_{m,2}H_{2} + b_{m,3}H_{3}\right)q_{m}$
 $H^{A} = a_{m+1,1}H_{1} + a_{m+1,2}H_{2} + a_{m+1,3}H_{3}$
enddo
 $a_{m+1} = a_{m} + \tau H^{A}p_{m}$

 $q_{m+1} = q_m + \tau H^A p_m$ $u_K = q_{m+1} + i p_m$ (17) Recall that the exact solution operator is time-symmetric, i.e., it satisfies $K(t + \tau, -\tau) \equiv K(t, \tau)^{-1}$. For this reason, we will focus only on schemes (15) and (16) satisfying the same symmetry, that is, $\tilde{K}(t + \tau, -\tau) = \tilde{K}(t, \tau)^{-1}$. This is automatically achieved if the coefficients verify the following conditions:

$$c_{k-j+1} = 1 - c_j, \quad a_{m+2-i,k+1-j} = a_{i,j}, \quad b_{m+1-i,k+1-j} = b_{i,j}$$
(18)

for j = 1, ..., k, i = 1, 2, ..., m. The integration schemes (15) and (16) are of order *r* if

$$\widetilde{K}(t,\tau) - K(t,\tau) = \mathcal{O}(\tau^{r+1}) \quad \text{as} \quad \tau \to 0.$$
 (19)

As is well known, time-symmetric integration schemes are of even order.

C. New schemes of order 4 and 6

We present next several time-symmetric schemes (15) and (16) of orders 4 and 6. In all of them, n = 3 and the coefficients c_j , j = 1, 2, 3, correspond to nodes of the Gauss-Legendre (GL) quadrature rule of order 6, that is,

$$c_1 = \frac{1}{2} - \frac{\sqrt{15}}{10}, \qquad c_2 = \frac{1}{2}, \qquad c_3 = \frac{1}{2} + \frac{\sqrt{15}}{10}.$$
 (20)

The remaining coefficients $\{a_{ij}, b_{ij}\}\$ must be determined to guarantee that (19) holds for r = 6. As a matter of fact, condition (19) translates into a set of polynomial equations on the coefficients (the so-called *order conditions*), whose solutions provide actual integration schemes. These equations have been obtained in Ref. 20 and are collected, for completeness, in a more simplified but equivalent form in Subsection 1 of the Appendix.

The process of obtaining appropriate values for the coefficients a_{ij} , b_{ij} is made simpler by noticing that the schemes (15) and (16) reduce in fact to (12) when applied to a constant matrix $H(t) = \tilde{H}$ as long as

$$a_i = \sum_{j=1}^n a_{i,j}, \qquad b_i = \sum_{j=1}^n b_{i,j}.$$
 (21)

It is then clear that a necessary condition for schemes (15) and (16), to be of order r for arbitrary smooth matrices H(t), is that (12) with coefficients (21) be an approximation of order r to the solution operator of system (10). In the particular case of 6th-order time-symmetric methods, such conditions are the first six order conditions listed in Subsection 1 of the Appendix, which only depend on the coefficients a_i , b_i . Hence, a convenient starting point in the construction process of 6th-order integrators of the form (15) and (16) for Equation (14) is to design previously a symplectic splitting method of the form (12) for the corresponding autonomous problem (10). In other words, to determine appropriate values for the coefficients (11) satisfying the first six order conditions given in Subsection 1 of the Appendix.

Once this is done, one has to solve the remaining 18 order conditions and (21) for the coefficients $a_{i,j}$, $b_{i,j}$. We have explored compositions (12) ranging from 8 to 24 stages and different orders, and tried to obtain time-symmetric schemes (15) and (16) based on them. In many cases, the remaining order conditions cannot be solved for real valued coefficients $a_{i,j}$ and $b_{i,j}$ satisfying (21). In particular, all the 6th-order schemes (12)

with m = 10 stages we have analyzed lead to complex coefficients $a_{i,j}$ and $b_{i,j}$. The most efficient schemes we have found are time-symmetric compositions of order 6 involving m = 11 stages. We have chosen as a representative the following:

- An 11-stage scheme, denoted in the sequel as SM^[6]₁₁.
- A second 11-stage scheme whose coefficients provide in the autonomous case an 8th-order approximation. It will be denoted as SM^[8]₁₁.

We have constructed other compositions with more stages, but they turn out to be clearly less efficient than $SM_{11}^{[6]}$ and $SM_{11}^{[8]}$ on all the numerical tests (not reported here) we have carried out with them. For completeness, we have also considered 8-stage schemes (15) and (16) of order 4 that is of order 6 for constant $H(t) = \tilde{H}$, which we denote as $SM_8^{[4]}$. The coefficients of the three proposed integrators are collected in Subsection 2 of the Appendix. The same coefficients with more significant digits as well as their implementation in several examples can be found in Ref. 30.

Although only schemes with coefficients c_j , j = 1, 2, 3 corresponding to the nodes (20) of the 6th-order Gauss-Legendre quadrature rule, are considered here, other choices are of course possible. In Subsection 3 of the Appendix, we provide the necessary transformations to adapt the present schemes to any other quadrature rule of order $r \ge 6$.

It is worth remarking that the *m*-stage splitting schemes presented here are particularly advantageous with respect to other time integrators (such as those considered in Section III) when the computational cost of evaluating the product $H(t_1)u$ is essentially the same as the cost of the product $(H(t_1))$ + $H(t_2)$)*u*. This is certainly the case when the Hamiltonian is of the form

$$H(t) = \sum_{i=1}^{s} f_i(t) H_i$$

with H_i constant and $s \ll N$, as occurs in many applications. Notice that the algorithm requires 2m products with real vectors, and this is equivalent to m products with complex vectors (and so roughly the same cost as approximating the action of the exponential by a polynomial of degree m). In addition, only one new real vector, v, needs to be kept in memory.

III. NUMERICAL ILLUSTRATIONS

We now illustrate the performance of the proposed methods on a pair of academic examples, namely, a high dimensional generalization of the Rosen–Zener model and the well-known Walker–Preston of a diatomic molecule. Although admittedly simple, they nevertheless represent adequately many typical applications and thus may serve as an indicative test bench of the methods proposed in this paper.

In our experiments, we compare with other well known schemes adapted to the linear problem (14). In particular, and guided by the conclusions of the analysis carried out in Ref. 2, we consider some highly efficient symplectic partitioned Runge-Kutta (splitting) methods. For our first example, we also include the efficiency diagrams obtained with a pair of explicit Runge–Kutta schemes as a standard reference. Specifically, the following integrators are taken for the numerical experiments:

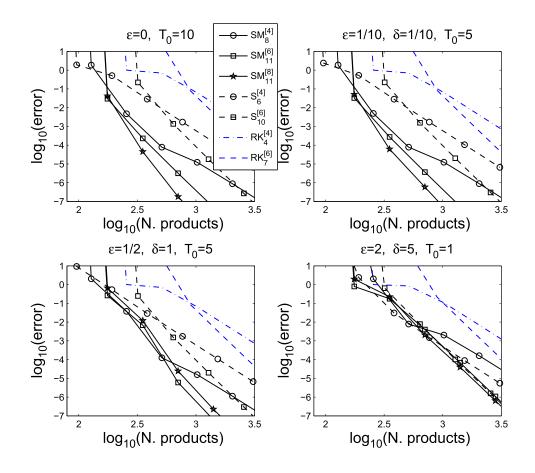


FIG. 1. Two-norm error in the evolution operator at the final time for the high dimensional Rosen–Zener model (23) in double logarithmic scale.

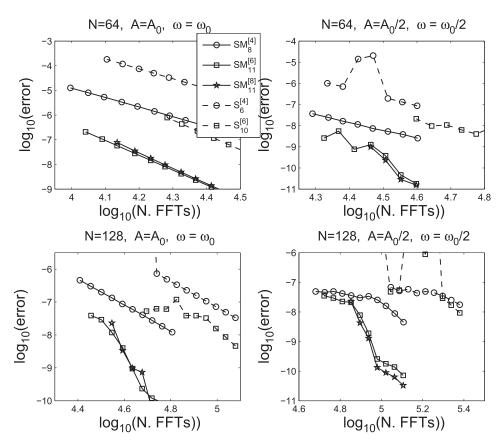


FIG. 2. Two-norm error in the vector solution of the discretized onedimensional SE (24) versus the number of FFT calls in double logarithmic scale.

- $RK_4^{[4]}$: The standard 4-stage 4th-order explicit Runge-Kutta method.
- RK₇^[6]: The 7-stage 6th-order explicit RK method that uses the Lobatto quadrature rule at internal stages to advance the time.
- $S_6^{[4]}$ and $S_{10}^{[6]}$: The 6-stage 4th-order and the 10-stage 6th-order splitting methods, respectively, designed in Ref. 31 for general separable systems and adapted to the present problem by considering the time as two new dependent variables as proposed in Ref. 13. In this way, the linear problem (14) is transformed into a nonlinear autonomous system. One step of these schemes is evaluated as follows:

do
$$i = 1, m$$

 $q_i = q_{i-1} + \tau a_i H(t_k + e_i \tau)$
 $p_i = p_{i-1} - \tau b_i H(t_k + d_i \tau) q_i$
(22)
enddo

with $d_i = \sum_{j=1}^{i} a_j$, $e_i = \sum_{j=0}^{i-1} b_j$ and $b_0 = 0$. These methods are independent of the Hamiltonian structure and have been tested in Ref. 2, showing a high performance.

• The new schemes SM^[4]₈ (order 4), SM^[6]₁₁, and SM^[8]₁₁ (order 6).

In all cases, the computational cost of the algorithms is estimated as the number of products required to produce the final result.

A. A high dimensional Rosen–Zener model

For our first illustration, we consider a generalization of the well known Rosen–Zenner model for a quantum system of two levels³² closely related to the model studied in Ref. 33. Specifically, the system is directly formulated as (4) with

$$H(t) = \omega(t) \sigma_3 \otimes I_k + V(t) \sigma_1 \otimes D^{[k]}.$$
 (23)

Here I_k is the $k \times k$ identity matrix, $D^{[k]}$ is a tridiagonal symmetric matrix with entries $D_{i,i}^{[k]} = 0$, $D_{i+1,i}^{[k]} = D_{i,i+1}^{[k]} = 1$, i = 1, ..., k - 1, $D_{k,k}^{[k]} = 0$, $\omega(t)$ is a time-dependent function, and σ_i are the Pauli matrices.

We take N = 2k = 20, $\omega(t) = w_0 + \varepsilon \cos(\delta t)$, $V(t) = V_0/\cosh(t/T_0)$, and $w_0 = 5$, $V_0 = 1/2$ and integrate along the interval $t \in [t_0, t_f]$ with $t_0 = -2$ and $t_f = t_0 + 8\pi$. We study the performance of the methods in the following cases:

- (i) $\varepsilon = 0, T_0 = 10.$
- (ii) $\varepsilon = \frac{1}{10}, \ \delta = 1/10, \ T_0 = 5.$
- (iii) $\varepsilon = \frac{1}{2}, \ \delta = 1, \ T_0 = 5.$
- (iv) $\varepsilon = \hat{2}, \ \delta = 5, \ T_0 = 1.$

The results obtained are shown in Figure 1. We observe that the new methods show the best performance and their superiority increases when the explicitly time-dependent functions are smooth. In that case, the main error originates from terms associated with the splitting method for the autonomous case, and the new methods are optimized for this case.

B. The Walker-Preston model

This constitutes a standard model for a diatomic molecule in a strong laser field.³⁴ The system is described by the onedimensional Schrödinger equation (in units such that $\hbar = 1$)

$$\frac{\partial}{\partial t}\psi(x,t) = \left(-\frac{1}{2\mu}\frac{\partial^2}{\partial x^2} + V(x) + f(t)x\right)\psi(x,t)$$
(24)

with $\psi(x, 0) = \psi_0(x)$. Here $V(x) = D(1 - e^{-\alpha x})^2$ is the Morse potential and $f(t)x = A\cos(\omega(t))x$ accounts for the laser field.

As explained in the Introduction, the usual procedure in its numerical treatment consists of defining the wave function ψ in a certain domain $x \in [x_0, x_N]$ that is subdivided into N parts of length $\Delta x = (x_N - x_0)/N$ with $x_i = x_0 + i\Delta x$, and then the vector $u(t) \in \mathbb{C}^N$ with components $u_i = (\Delta x)^{1/2} \psi(x_{i-1}, t)$, i = 1, ..., Nis formed. Notice that the 2-norm ||u(t)|| remains constant for all t and, in order to use the Fast Fourier Transform (FFT) algorithm, periodic boundary conditions $\psi(x_0, t) = \psi(x_N, t)$ are assumed.

Here we take the interval $x \in [-0.8, 4.32]$, subdivided into N = 64 and N = 128 parts of length $\Delta x = 0.08$ and $\Delta x = 0.04$, respectively, and the parameters are chosen as $\mu = 1745$ a.u., D = 0.2251 a.u., and $\alpha = 1.1741$ a.u. (corresponding to the HF molecule). For each choice of N, we take for time-dependent interaction $A = A_0 = 0.011025$ a.u. and laser frequency $\omega = \omega_0 = 0.01787$. The numerical experiments are then repeated with $A = A_0/2$ and $\omega = \omega_0/2$, corresponding to a reduction in the intensity and frequency of the laser. As initial conditions we take the ground state of the Morse potential,

$$\phi(x) = \sigma \exp\left(-(\gamma - 1/2)\alpha x\right) \exp\left(-\gamma e^{-\alpha x}\right),$$

with $\gamma = 2D/w_0$, $w_0 = \alpha \sqrt{2D/\mu}$, and σ is a normalizing constant, and integrate over the time interval $t \in [0, 10 T_0]$ with $T_0 = 2\pi/\omega$ (the choice $\omega = \omega_0/2$ requires a twice longer time integration).

To check accuracy, we measure the two-norm error in the solution at the final time, $u(10T_0)$. As usual, the exact solution is accurately approximated using a sufficiently small time step.

Figure 2 shows the efficiency plots for the methods. The largest time step (i.e., the smaller number of FFTs) corresponds to the stability limit of the method (an overflow or an exceedingly large error appears when the time step is slightly increased). We observe that the relative performance of the new methods increases both when taking a finer mesh or when the time-dependent interaction is weaker and smoother. We observe that in such limit, the 6th-order method that corresponds to an 8th-order one in the autonomous case shows the best performance when high accuracies are desired. The 4th-order method shows less accuracy but allows for longer time steps and more frequent outputs.

IV. CONCLUDING REMARKS

We have presented several symplectic splitting methods for the numerical integration in time of the Schrödinger equation previously discretized in space when the Hamiltonian is an explicitly time-dependent real operator. Our approach differs from others existing in the literature in that we do

not need to impose beforehand a slow time-dependence in the potential so that it can be taken as approximately constant in each time subinterval along the integration. It is based on the fact that the semidiscretized equation can be formulated as a classical linear Hamiltonian system for which the solution operator can be approximated as a product of matrices involving linear combinations of the (semidiscretized) time-dependent Hamiltonian on quadrature nodes at each time step. Here we have considered methods of orders four and six based on the Gauss-Legendre quadrature rule of order six, but the extension to other quadrature rules (as shown in Subsection 3 of the Appendix) is rather straightforward. Although originally intended for the numerical integration of the Schrödinger equation with a time-dependent potential, the new methods are still valid for arbitrary linear ODE systems of the form $\dot{u} = -iH(t)u$ for an arbitrary real matrix H(t)provided its time dependence on t is reasonably smooth. Moreover, the proposed method could also be used to solve systems like x' = M(t)y, y' = N(t)x arising, e.g., in elastic waves analysis.35

ACKNOWLEDGMENTS

The authors acknowledge Ministerio de Economía y Competitividad (Spain) for financial support through Project Nos. MTM2013-46553-C3 and MTM2016-77660-P (AEI/FEDER, UE). Additionally, A.M. has been partially supported by the Basque Government (Consolidated Research Group No. IT649-13).

APPENDIX: ON THE CONSTRUCTION OF METHODS

1. Order conditions

To derive the order conditions for the schemes (15) and (16) with n = 3 and the coefficients c_j given by (20), it is advantageous to first write (15) in terms of a new set of variables in order to avoid redundancies and limit the presence of contributions at orders higher than six. Specifically, instead of considering the matrices $H_i \equiv H(t + c_i\tau)$, i = 1, 2, 3, so that $\tau H_i = O(\tau)$, we introduce the variables

$$\begin{aligned} \gamma_1 &\equiv \tau H_2 = \mathcal{O}(\tau), \\ \gamma_2 &\equiv \frac{\sqrt{15}}{3} \tau (H_3 - H_1) = \mathcal{O}(\tau^2), \\ \gamma_3 &\equiv \frac{10}{3} \tau (H_1 - 2H_2 + H_3) = \mathcal{O}(\tau^3). \end{aligned}$$
(A1)

As a matter of fact, the γ_i are the (i-1)th derivatives, evaluated at the midpoint $t + \tau/2$ and divided by (i-1)!, of a matrix that interpolates H(t) at the nodes. Then, schemes (15) and (16) can be expressed as

$$\widetilde{K}(t_k, \tau) = \prod_{i=1}^{m+1} \begin{pmatrix} I & 0\\ -\tau(b_{i,1}H_1 + b_{i,2}H_2 + b_{i,3}H_3) I \end{pmatrix} \begin{pmatrix} I \ \tau(a_{i,1}H_1 + a_{i,2}H_2 + a_{i,3}H_3) \\ 0 & I \end{pmatrix}$$
$$= \prod_{i=1}^{m+1} \begin{pmatrix} I & 0\\ -(b_i\gamma_1 + b'_i\gamma_2 + b''_i\gamma_3) I \end{pmatrix} \begin{pmatrix} I \ (a_i\gamma_1 + a'_i\gamma_2 + a''_i\gamma_3) \\ 0 & I \end{pmatrix},$$
(A2)

with $b_{m+1,j} = b_{m+1} = b'_{m+1} = 0$, j = 1, 2, 3. In the autonomous case, we have that $\gamma_1 = \tau \tilde{H}$, $\gamma_2 = \gamma_3 = 0$ and the composition (12) is recovered. If the composition satisfies the symmetry conditions (18), the scheme will be of order 6 when applied to a system of the form (14) whenever the following conditions are satisfied (see Ref. 20 for the derivation of equivalent conditions for symmetric schemes (12) of order 6):

$$\begin{split} &\sum_{i=1}^{m+1} a_i = 1, \qquad \sum_{i=1}^m b_i = 1, \\ &\sum_{m+1 \ge i_3 > i_2 > i_1 \ge 1} a_{i_3} b_{i_2} a_{i_1} = \frac{1}{6}, \qquad \sum_{m \ge i_3 > i_2 > i_1 \ge 1} b_{i_3} a_{i_2} b_{i_1} = \frac{1}{6}, \\ &\sum_{m+1 \ge i_3 > i_2 > i_1 \ge 1} a_{i_3} b_{i_4} a_{i_3} b_{i_2} a_{i_1} = \frac{1}{120}, \\ &\sum_{m \ge i_3 > i_2 > i_1 \ge 1} b_{i_5} a_{i_4} b_{i_3} a_{i_2} b_{i_1} = \frac{1}{120}, \\ &\sum_{m \ge i_3 > i_2 > i_1 \ge 1} a_{i_2}' b_{i_1} = \frac{1}{12}, \qquad \sum_{m \ge i_2 > i_1 \ge 1} b_{i_2}' a_{i_1} = \frac{1}{12}, \\ &\sum_{m+1 \ge i_2 > i_1 \ge 1} a_{i_2}' b_{i_1} = -\frac{1}{12}, \qquad \sum_{m \ge i_2 > i_1 \ge 1} b_{i_2}' a_{i_1} = -\frac{1}{120}, \\ &\sum_{m+1 \ge i_2 > i_1 \ge 1} a_{i_2}' b_{i_1}' = -\frac{1}{240}, \qquad \sum_{m \ge i_2 > i_1 \ge 1} b_{i_2}' a_{i_1}' = -\frac{1}{240}, \\ &\sum_{m+1 \ge i_3 > i_2 > i_1 \ge 1} a_{i_3} b_{i_2} a_{i_1}' = \frac{1}{120}, \qquad \sum_{m \ge i_2 > i_1 \ge 1} b_{i_3} a_{i_2}' b_{i_1}' = \frac{1}{120}, \\ &\sum_{m+1 \ge i_3 > i_2 > i_1 \ge 1} a_{i_3} b_{i_2} a_{i_1}' = \frac{1}{60}, \qquad \sum_{m \ge i_3 > i_2 > i_1 \ge 1} b_{i_3} a_{i_2}' b_{i_1}' = \frac{1}{120}, \\ &\sum_{m+1 \ge i_3 > i_2 > i_1 \ge 1} a_{i_3} b_{i_2} a_{i_1}' = \frac{1}{240}, \qquad \sum_{m \ge i_3 > i_2 > i_1 \ge 1} b_{i_3} a_{i_2}' b_{i_1}' = \frac{1}{120}, \\ &\sum_{m+1 \ge i_3 > i_2 > i_1 \ge 1} a_{i_3} b_{i_2} a_{i_1}' = \frac{1}{240}, \qquad \sum_{m \ge i_3 > i_2 > i_1 \ge 1} b_{i_3} a_{i_2}' b_{i_1}' = \frac{1}{120}, \\ &\sum_{m+1 \ge i_3 > i_2 > i_1 \ge 1} a_{i_3} b_{i_2} a_{i_1}' = -\frac{1}{120}, \qquad \sum_{m \ge i_3 > i_2 > i_1 \ge 1} b_{i_3} a_{i_2} b_{i_1}' = -\frac{1}{120}, \\ &\sum_{m+1 \ge i_3 > i_2 > i_1 \ge 1} a_{i_3} b_{i_2} a_{i_1}' = -\frac{1}{120}, \qquad \sum_{m \ge i_3 > i_2 > i_1 \ge 1} b_{i_3} a_{i_2} b_{i_1}' = -\frac{1}{120}, \\ &\sum_{m+1 \ge i_3 > i_2 > i_1 \ge 1} a_{i_3} b_{i_2} a_{i_1}' = -\frac{1}{120}, \qquad \sum_{m \ge i_3 > i_2 > i_1 \ge 1} b_{i_3} a_{i_2} b_{i_1}' = -\frac{1}{120}, \\ &\sum_{m+1 \ge i_3 > i_2 > i_1 \ge 1} a_{i_3} b_{i_2} a_{i_1}' = -\frac{1}{120}, \qquad \sum_{m \ge i_3 > i_2 > i_1 \ge 1} b_{i_4} a_{i_3} b_{i_2} a_{i_1}' = -\frac{1}{80}, \\ &\sum_{m+1 \ge i_3 > i_2 > i_1 \ge 1} a_{i_4} b_{i_3} a_{i_2}' b_{i_1}' = -\frac{1}{240}. \end{aligned}$$

Here, from (A1) and (A2), one has

$$\begin{pmatrix} a_{i,1} \ b_{i,1} \\ a_{i,2} \ b_{i,2} \\ a_{i,3} \ b_{i,3} \end{pmatrix} = \begin{pmatrix} 0 - \sqrt{\frac{5}{3}} & \frac{10}{3} \\ 1 & 0 & -\frac{20}{3} \\ 0 & \sqrt{\frac{5}{3}} & \frac{10}{3} \end{pmatrix} \begin{pmatrix} a_i \ b_i \\ a'_i \ b'_i \\ a''_i \ b''_i \end{pmatrix}.$$
(A3)

2. Coefficients of the methods

Next we collect the coefficients of the symplectic splitting methods proposed in this work for the numerical integration of Equation (14). For the sake of clarity, we write them in matrix form. The same coefficients, with more digits, are collected in Ref. 30. All these coefficients have to be supplemented by the nodes c_j given in (20).

a. Method $SM_8^{[4]}$

$$a = \begin{pmatrix} 0.056\,543\,643\,80 & 0.013\,657\,706\,809 & -0.034\,367\,547\,779 \\ 0.151\,876\,511\,53 & -0.066\,217\,362\,266 & 0.075\,469\,735\,351 \\ 0.074\,446\,942\,50 & 0.208\,318\,930\,216 & -0.042\,763\,740\,386 \\ -0.011\,712\,456\,09 & -0.002\,171\,489\,464 & 0.008\,284\,688\,848 \\ 0 & 0.137\,268\,873\,853 & 0 \\ a_{4,3} & a_{4,2} & a_{4,1} \\ a_{3,3} & a_{3,2} & a_{3,1} \\ a_{2,3} & a_{2,2} & a_{2,1} \\ a_{1,3} & a_{1,2} & a_{1,1} \end{pmatrix},$$

•

	0.109 165 185 01	0.013 444 365 00	-0.019 256 622 788
	0.188 194 699 07	-0.03970769739	0.058 395 655 885
	0.141 154 416 25	0.570 642 655 82	-0.043 047 573 981
1.	-0.100 060 196 70	-0.322 157 101 21	-0.056767784980
<i>b</i> =	$b_{4,3}$	$b_{4,2}$	$b_{4,1}$
	$b_{3,3}$	$b_{3,2}$	$b_{3,1}$
	$b_{2,3}$	$b_{2,2}$	$b_{2,1}$
	$b_{1,3}$	$b_{1,2}$	$b_{1,1}$)

b. Method $SM_{11}^{[6]}$

	(0.064 229 455 56	-0.025 341 566 51	0.007 599 565 74	`
	-0.044 424 869 66	-0.02451080257	0.008 244 001 06	
	0.205 475 526 18	0.021 439 623 03	-0.00844862275	
	0.066 283 190 90	0.122 950 863 12	-0.02118047454	
	-0.003 551 371 84	0.314 392 364 17	$-a_{4,1}$	
<i>a</i> –	0.003 488 712 73	-0.18670825374	$-a_{5,1}$	
<i>a</i> =	<i>a</i> _{6,3}	$a_{6,2}$	$a_{6,1}$	'
	<i>a</i> _{5,3}	$a_{5,2}$	$a_{5,1}$	
	<i>a</i> _{4,3}	$a_{4,2}$	$a_{4,1}$	
	<i>a</i> _{3,3}	<i>a</i> _{3,2}	$a_{3,1}$	
	<i>a</i> _{2,3}	$a_{2,2}$	$a_{2,1}$	
	$(a_{1,3})$	$a_{1,2}$	$a_{1,1}$)	/

	(0.198 931 884 48	-0.01661701661	0.002 015 615 63
	-0.030 836 241 53	-0.01190945158	0.001 688 789 82
	0.079 650 985 44	0.044 994 246 37	0.009 110 447 84
	0.082 864 339 33	0.186 548 251 04	-0.065 648 043 24
	0.012 909 944 48	-0.01176016691	$-b_{5,1}$
<i>b</i> =	0	0.061 932 719 82	0.
	$b_{5,3}$	$b_{5,2}$	$b_{5,1}$
	$b_{4,3}$	$b_{4,2}$	$b_{4,1}$
	$b_{3,3}$	$b_{3,2}$	$b_{3,1}$
	$b_{2,3}$	$b_{2,2}$	$b_{2,1}$
	<i>b</i> _{1,3}	$b_{1,2}$	$b_{1,1}$)

c. Method $SM_{11}^{[8]}$

	-0.100 359 598 31 0.217 485 576 77 0.083 044 197 90 0.007 699 808 87	-0.055 358 596 01 -0.159 313 968 90 0.209 628 140 57 0.104 202 478 18 0.290 721 833 65	0.045 184 103 41 -0.058 916 052 51 -0.019 749 426 81 0.097 739 158 98	
<i>a</i> =	0.0383/0/3411 $a_{6,3}$	-0.16765766527 $a_{6,2}$	$\begin{bmatrix} -0.12/8105/054\\a_{6.1} \end{bmatrix}$,	
	<i>a</i> _{5,3}	a _{5,2}	a _{5,1}	
	<i>a</i> _{4,3}	$a_{4,2}$	$a_{4,1}$	
	<i>a</i> _{3,3}	<i>a</i> _{3,2}	$a_{3,1}$	
	$a_{2,3}$	$a_{2,2}$	$a_{2,1}$	
	$(a_{1,3})$	$a_{1,2}$	$a_{1,1}$)	

	(0.154 171 842 09	-0.01842607545	0.007 131 251 77
	-0.006 347 779 03	-0.00389065405	0.001 097 338 89
	0.113 616 267 16	0.024 959 240 32	-0.013 594 387 63
	0.034 287 009 27	0.201 283 836 52	-0.008 359 528 72
	-0.000 119 805 05	-0.00852170559	$b_{5,1}$
<i>b</i> =	-0.003 984 625 91	0.053 635 160 96	$b_{6,1}$
	$b_{5,3}$	$b_{5,2}$	$b_{5,1}$
	$b_{4,3}$	$b_{4,2}$	$b_{4,1}$
	$b_{3,3}$	$b_{3,2}$	$b_{3,1}$
	$b_{2,3}$	$b_{2,2}$	$b_{2,1}$
	$b_{1,3}$	$b_{1,2}$	$b_{1,1}$)

3. Schemes with nodes of an arbitrary quadrature rule

Methods (15) and (16) can be used with any other quadrature rule of order $r \ge 6$, say $\{\hat{w}_i, \hat{c}_i\}_{i=1}^k$. All that is required is to replace in (17) the nodes c_i , i = 1, 2, 3 by \hat{c}_i , i = 1, ..., k, and the coefficients $\{a_{i,j}, b_{i,j}\}$ by the new $\{\hat{a}_{i,j}, \hat{b}_{i,j}\}$, so that the scheme reads

$$\widehat{K}(t_k,\tau) = \prod_{i=1}^{m+1} \begin{pmatrix} I & 0\\ -\tau \sum_{i=1}^k \hat{b}_{i,1} \hat{H}_i & I \end{pmatrix} \begin{pmatrix} I & \tau \sum_{i=1}^k \hat{a}_{i,1} \hat{H}_i \\ 0 & I \end{pmatrix}$$

with $\hat{H}_i = H(t + \hat{c}_i \tau)$, i = 1, ..., k. The new coefficients $\{\hat{a}_{i,j}, \hat{b}_{i,j}\}$ are chosen such that

$$\widehat{K}(t_k, \tau) = \widetilde{K}(t_k, \tau) + \mathcal{O}(\tau^7)$$

but this is satisfied if

$$\hat{a} = a Q^{-1} \widehat{Q}, \qquad \hat{b} = b Q^{-1} \widehat{Q}.$$
 (A4)

Here, $a \in \mathbb{R}^{(m+1)\times 3}$, $b \in \mathbb{R}^{m\times 3}$, $\hat{a} \in \mathbb{R}^{(m+1)\times k}$, $\hat{b} \in \mathbb{R}^{m\times k}$ denote the matrices with coefficients $a_{i,j}, b_{i,j}, \hat{a}_{i,j}, \hat{b}_{i,j}$, respectively, and the elements of the matrices $Q \in \mathbb{R}^{3\times 3}$, $\hat{Q} \in \mathbb{R}^{3\times k}$ are given by

$$Q_{i,j} = w_j \left(c_j - \frac{1}{2} \right)^{i-1}, \quad \widehat{Q}_{i,l} = \hat{w}_l \left(\hat{c}_k - \frac{1}{2} \right)^{i-1},$$

 $i, j = 1, 2, 3, l = 1, \dots, k$, with

$$(c_1, c_2, c_3) = \left(\frac{1}{2} - \frac{\sqrt{15}}{10}, \frac{1}{2}, \frac{1}{2} + \frac{\sqrt{15}}{10}\right),$$
$$(w_1, w_2, w_3) = \frac{1}{18}(5, 8, 5).$$

As an illustration, suppose H(t) is known only at equidistant times. Then we can use the 6h-order Newton–Cotes quadrature rule with $\hat{c}_i = (i - 1)/4$, i = 1, ..., 5,

$$(\hat{w}_1, \hat{w}_2, \hat{w}_3, \hat{w}_4, \hat{w}_5) = \frac{1}{90}(7, 32, 12, 32, 7),$$

so that

$$Q^{-1}\widehat{Q} = \begin{pmatrix} \frac{7}{150}(5+\sqrt{15}) & \frac{4}{75}(5+2\sqrt{15}) & 0 & \frac{4}{75}(5-2\sqrt{15}) & \frac{7}{150}(5-\sqrt{15}) \\ -\frac{7}{60} & \frac{7}{15} & \frac{3}{10} & \frac{7}{15} & -\frac{7}{60} \\ \frac{7}{150}(5-\sqrt{15}) & \frac{4}{75}(5-2\sqrt{15}) & 0 & \frac{4}{75}(5+2\sqrt{15}) & \frac{7}{150}(5+\sqrt{15}) \end{pmatrix}$$

Then the corresponding coefficients for schemes $SM_8^{[4]}$, $SM_{11}^{[6]}$, and $SM_{11}^{[8]}$ can be obtained (from those collected in Subsection 2 of the Appendix) by (A4).

We have carried out the experiments of Section III with the present symplectic splitting methods but using different quadrature rules. Our experiments (not reported here) show that the plots are essentially identical when considering the equidistant 6th-order Newton-Cotes or the 8th-order GL quadrature rule. If, on the other hand, the 4th-order GL quadrature rule is used instead, the new 4th-order method remains basically the same, but the 6th-order methods lose some accuracy, as expected.

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