# Global integration of the Schrödinger equation: a short iterative scheme within the wave operator formalism using discrete Fourier transforms 

Georges Jolicard, Arnaud Leclerc

## - To cite this version:

Georges Jolicard, Arnaud Leclerc. Global integration of the Schrödinger equation: a short iterative scheme within the wave operator formalism using discrete Fourier transforms. Journal of Physics A: Mathematical and Theoretical, 2015, 48 (22), pp.225205. 10.1088/1751-8113/48/22/225205 . hal02641194v2

HAL Id: hal-02641194
https://hal.science/hal-02641194v2
Submitted on 14 Jun 2022

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire HAL, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

# Global integration of the Schrödinger equation: a short iterative scheme within the wave operator formalism using discrete Fourier transforms 

Arnaud Leclerc§, Georges Jolicard $\dagger$<br>§ SRSMC, UMR CNRS 7565, Université de Lorraine, 1Bd Arago 57070 Metz, France<br>† Institut Utinam, UMR CNRS 6213, Université de Franche-Comté,<br>Observatoire de Besançon, 41 bis avenue de l'Observatoire, BP1615, 25010<br>Besançon Cedex, France<br>E-mail: arnaud.leclerc@univ-lorraine.fr, georges.jolicard@utinam.cnrs.fr


#### Abstract

. A global solution of the Schrödinger equation for explicitly time-dependent Hamiltonians is derived by integrating the non-linear differential equation associated with the time-dependent wave operator. A fast iterative solution method is proposed in which, however, numerous integrals over time have to be evaluated. This internal work is done using a numerical integrator based on Fast Fourier Transforms (FFT). The case of a transition between two potential wells of a model molecule driven by intense laser pulses is used as an illustrative example. This application reveals some interesting features of the integration technique. Each iteration provides a global approximate solution on grid points regularly distributed over the full time propagation interval. Inside the convergence radius, the complete integration is competitive with standard algorithms, especially when high accuracy is required.


## 1. Introduction

The numerical solution of the Schrödinger equation $i \hbar \partial \Psi / \partial t=H \Psi$ plays a key role in the understanding of molecular dynamics processes. Molecular inelastic [1] and reactive collisions [2] have long been treated using quantal wave packet methods. In procedures such as the semiclassical treatment of multiphoton processes [3] explicitly time-dependent Hamiltonians have to be used to describe the molecule-field interaction. A popular and efficient wave-packet propagation algorithm is the MultiConfiguration Time-Dependent Hartree method, which is able to handle a large variety of high-dimensional problems [4]. Here, our aim is not to deal with many dimensions but to develop an efficient way of treating problems with rapidly oscillating time dependencies in the Hamiltonian operator or in the wavefunctions. Several algorithms which are used for time-resolved experiments with time-independent Hamiltonians can be adapted to the time-dependent case and are described in [5] and references therein. However for such methods the length of the integration steps must be reduced to handle the high frequency terms. The propagation scheme is then based on the decomposition of the evolution operator using small time increments.

$$
\begin{equation*}
U(t, 0)=\prod_{n=0}^{N-1} U((n+1) \Delta t, n \Delta t) \tag{1}
\end{equation*}
$$

where $\Delta t=t / N$ and

$$
\begin{equation*}
U(t+\Delta t, t) \simeq \exp [-(i / \hbar) H(t+\Delta t / 2) \Delta t] \tag{2}
\end{equation*}
$$

These step by step integration schemes will be referred to in later sections as "continuous methods". Second order differencing schemes (SOD) [6], split operator methods [7] and the short iterative Lanczos propagation [8] all give cumulative propagation errors which are proportional to $(\Delta t)^{3}$ in (1) and also errors due to the approximate calculation of the action of $\exp [-(i / \hbar) H \Delta t]$ on the wave function. Other high-accuracy integrators which can be used for multi-dimensional systems include the generalization of the sympletic partitioned Runge-Kutta method developed by Sanz-Serna and Portillo [9] and the method proposed by Kormann [10] in which the Hamiltonian $H$ is replaced by a suitable truncation of the Magnus series. Unfortunately, these continuous integrators suffer from limitations which could prevent their use in some cases. SOD cannot handle non-Hermitian Hamiltonians, while the split operator scheme requires that the kinetic operator does not mix coordinates and their associated momenta. The calculation of the Magnus development is only tractable if the couplings have separate time and coordinate dependencies [10].

A recently proposed global treatment, the Constrained Adiabatic Trajectory Method (CATM), [11, 12, 13, 14] introduces the Floquet Hamiltonian $H_{F}=H-$ $i \hbar \partial / \partial t$ defined in an extended Hilbert space including the time coordinate. The dynamical Schrödinger equation is transformed into an equivalent partial eigenvalue problem in the extended space and the wave function is forced to conform to consistent boundary conditions over a short artificial time extension by using a time-dependent absorbing potential. In [14], the CATM was formulated as a global integrator of the Schrödinger equation which does not belong to the category of methods described by (1) and (2) and it was concluded that the CATM is well suited to the description of systems driven by Hamiltonians with explicit and complicated time variations. The method does not suffer from cumulative errors (because of its global character) and the only error sources are the non-completeness of the finite molecular and temporal
basis sets used and the imperfection of the time-dependent absorbing potential. The dynamics of a quantum system can also be obtained in a global way (as in the CATM) by using the time-dependent wave operator concept (TDWO) [15]. In this formalism the dynamics is formally separated into a simple evolution within a given active subspace driven by an effective low-dimensional Hamiltonian plus a secondary evolution from the subspace to the complete space. The TDWO concept has also been generalized [16] to treat almost adiabatic quantum dynamics. This generalization is based on a time-dependent adiabatic deformation of the active space and is useful for dynamics which do not escape too far from an adiabatic subspace. However the effectiveness of both TDWO and CATM hinges on the capacity to integrate some nonlinear differential equations and this is the central problem treated in this work. A solution of this problem is essential to preserve the principal advantage of these formalisms, i.e. their capacity to give a global solution over the whole interaction time. To reach this goal an iterative solution seems appropriate. Such a solution was proposed in ref. [17], using a time-dependent version of the Recursive Distorted Wave Approximation (RDWA) [15] together with non-linear transformations such as Padé approximants [18]. Nevertheless, this iterative solution is not fully satisfactory and often fails because of a small radius of convergence and a high sensitivity to the choice of the initial guessed solution.

The present paper proposes a new iterative solution of the time-dependent wave operator equations. This is equivalent to solving the time-dependent Schrödinger equation without using the approximation of (1) and (2) at any stage of the calculation. In section 2 the basic wave operator equations are presented together with the iterative integration procedure. In practice we limit ourselves to the non-degenerate case (one-dimensional active subspace) which describes the dynamics issuing from a given quantum state. This is equivalent to solving the Schrödinger equation for the wavefunction. The iteration procedure involves calculating numerous integrals over time of matrix and vector elements. Section 3 explores the possibility of using Fast Fourier Transforms (FFT) to compute these integrals. In section 4 we illustrate the algorithm by studying the transition of a vibrational wave packet between the two wells of a model potential energy surface under the influence of two laser fields. Various numerical features are analysed in details. Section 5 gives a discussion and conclusion.

## 2. Iterative calculation of the time dependent wave operator

### 2.1. Wave operator equations

Let $\mathcal{H}$ be the Hilbert space associated with a molecular system and let $S_{o}$ be a model subspace of rank $m$. It is assumed that the initial molecular state is included within $S_{o}$. The orthogonal projector corresponding to the model space is called $P_{o}$, with $P_{o}^{2}=P_{o}, P_{o}^{\dagger}=P_{o}, \operatorname{tr}\left(P_{o}\right)=m$. During the evolution, the model space is continuously transformed into successive active spaces $S(t)$ whose projectors $P(t)$ are solutions of the Schrödinger-von Neumann equation:

$$
\begin{equation*}
i \hbar \frac{\partial P(t)}{\partial t}=[H(t), P(t)] \tag{3}
\end{equation*}
$$

The wave operator associated with the two subspaces $S_{o}$ and $S(t)$ is defined as [15]

$$
\begin{equation*}
\Omega(t)=P(t)\left(P_{o} P(t) P_{o}\right)^{-1}=U(t, 0 ; H)\left(P_{o} U(t, 0 ; H) P_{o}\right)^{-1} \tag{4}
\end{equation*}
$$

where $U$ represents the quantum evolution operator associated with the Hamiltonian $H(t)$, i.e. $i \hbar \frac{\partial U(t, 0 ; H)}{\partial t}=H(t) U(t, 0 ; H)$. In (4), $\left(P_{o} P P_{o}\right)^{-1}$ is the inverse of $P$ within $S_{o}$. It exists only if the Fubini-Study distance between $P_{o}$ and $P$ is small: $\operatorname{dist}_{F S}\left(P_{o}, P\right) \leq \frac{\pi}{2}$ [16], i.e. if the dynamics does not escape too far from the initial subspace. In this framework the time evolution can be written as

$$
\left\{\begin{array}{l}
U(t, 0 ; H) P_{o}=\Omega(t) U\left(t, 0 ; H_{e f f}\right)  \tag{5}\\
H_{e f f}(t)=P_{o} H(t) \Omega(t)
\end{array}\right.
$$

In equation 5 the evolution issuing from the initial state is separated into two terms. The first one, $U\left(t, 0 ; H_{e f f}\right)$, describes the dynamics within the model space $S_{o}$. The second one can be written as $\Omega(t)=P_{o}+X(t)$ and it induces transitions from the model space to the full space, the off diagonal part ( $X=Q_{o} \Omega P_{o}$ ) inducing transitions to the complementary space exclusively ( $Q_{o}$ being the projector associated with the complementary space). It is evident that the factorisation of $U\left(H_{e f f}\right)$, which includes all the fast variations within $S_{o}$, is an advantage for the dynamical integration process if the model space is well chosen. It can be proven that the reduced wave operator $X(t)=Q_{o} \Omega P_{o}$ satisfies [15]

$$
\begin{equation*}
i \hbar \frac{\partial X(t)}{\partial t}=Q_{o}(1-X(t)) H(t)(1+X(t)) P_{o} \tag{6}
\end{equation*}
$$

Solving this equation is the central point of the wave operator formalism. The solution gives $\Omega(t)$ as well as $H_{e f f}(t)$ (cf equation 5), from which a partial evolution operator can be obtained.

A continuous integration of this non-linear differential equation has been proposed in section VI of the review [15]. Unfortunately, many approximations were necessary to make the calculation tractable and the final result was not satisfactory in the strong coupling regime. To derive a new global solution, we can use the fact that $Q_{o}(1-X) \frac{\partial}{\partial t}(1+X) P_{o}=Q_{o} \frac{\partial}{\partial t} X P_{o}$ to rewrite (6) as

$$
\begin{equation*}
Q_{o}(1-X(t)) H_{F}(t)(1+X(t)) P_{o}=0 \tag{7}
\end{equation*}
$$

where $H_{F}(t)$ is the Floquet Hamiltonian,

$$
\begin{equation*}
H_{F}(t)=H(t)-i \hbar \frac{\partial}{\partial t} . \tag{8}
\end{equation*}
$$

We now derive an iterative treatment, assuming that eq (7) is not perfectly satisfied at a finite iteration order $n$,

$$
\begin{equation*}
Q_{o}\left(1-X^{(n)}(t)\right) H_{F}(t)\left(1+X^{(n)}(t)\right) P_{o}=\Delta^{(n)}(t) \neq 0 \tag{9}
\end{equation*}
$$

We are looking for the increment $\delta X^{(n)}$ which exactly solves the problem:

$$
\begin{equation*}
Q_{o}\left(1-X^{(n)}(t)-\delta X^{(n)}(t)\right) H_{F}(t)\left(1+X^{(n)}(t)+\delta X^{(n)}(t)\right)=0 \tag{10}
\end{equation*}
$$

Expanding (10) gives four terms of different significance,

$$
\begin{align*}
& \Delta^{(n)}-Q_{o} \delta X^{(n)}(t) H_{F}(t) \delta X^{(n)}(t) P_{o} \\
& -Q_{o} \delta X^{(n)}(t) H_{F}(t)\left(1+X^{(n)}(t)\right) P_{o} \\
& +Q_{o}\left(1-X^{(n)}(t)\right) H_{F}(t) \delta X^{(n)}(t) P_{o}=0 \tag{11}
\end{align*}
$$

In (11), the first term is the error at the previous iteration defined in (9). The second term is quadratic in the increment $\left(\delta X^{(n)}(t)\right)$ and can be neglected; this is
an approximation commonly used in such iterative treatments. The third term can be rewritten as

$$
\begin{align*}
& Q_{o} \delta X^{(n)} H_{F}\left(1+X^{(n)}\right) P_{o} \\
= & Q_{o} \delta X^{(n)}\left[P_{o} H\left(P_{o}+Q_{o} X^{(n)} P_{o}\right)-P_{o} i \hbar \frac{\partial}{\partial t}\left(P_{o}+Q_{o} X^{(n)} P_{o}\right)\right] \\
= & Q_{o} \delta X^{(n)} P_{o} H\left(P_{o}+Q_{o} X^{(n)} P_{o}\right) \\
= & Q_{o} \delta X^{(n)}(t) H_{e f f}^{(n)}(t) P_{o}, \tag{12}
\end{align*}
$$

with

$$
\begin{equation*}
H_{e f f}^{(n)}(t)=P_{o} H(t)\left(P_{o}+X^{(n)}(t)\right) \tag{13}
\end{equation*}
$$

The last term in (11) needs more detailed attention:

$$
\begin{align*}
& Q_{o}\left(1-X^{(n)}(t)\right) H_{F} \delta X^{(n)}(t) P_{o} \\
= & Q_{o} H \delta X^{(n)} P_{o}-Q_{o} X^{(n)} H \delta X^{(n)} i \hbar \frac{\partial}{\partial t} Q_{o} X^{(n)} P_{o} \tag{14}
\end{align*}
$$

We cannot handle differential equations in which $\left(H \delta X^{(n)}\right)$ or $\left(X^{(n)} H\right)$ couple the unknown components of $\delta X^{(n)}$ to all the complementary space spanned by $Q_{o}$. We shall thus assume that the non-diagonal elements of $H$ in the complementary space, given by $Q_{o}\left(H-H_{\text {diag }}\right) Q_{o}$, are small and make two last approximations,

$$
\begin{align*}
& Q_{o} H(t) Q_{o} \approx Q_{o}[H(t)]_{\text {diag }} Q_{o} \\
& Q_{o} X^{(n)}(t) H(t) Q_{o} \approx Q_{o}\left[X^{(n)}(t) H(t)\right]_{\text {diag }} Q_{o} \tag{15}
\end{align*}
$$

This simplification is important but does not have important consequences if the procedure still converges. The neglected terms are gradually taken into account during the iterative process and finally equation (7) is satisfied. In summary, the increment defined in (10) can be computed by solving

$$
\begin{equation*}
i \hbar \frac{\partial}{\partial t} \delta X^{(n)}(t)=\Delta^{(n)}(t)-\delta X^{(n)}(t) H_{e f f}^{(n)}(t)+\tilde{H}_{\text {diag }}^{(n)}(t) \delta X^{(n)}(t) \tag{16}
\end{equation*}
$$

with $H_{\text {eff }}$ defined in (13) and

$$
\begin{equation*}
\tilde{H}_{d i a g}^{(n)}(t)=Q_{o}\left[H(t)-X^{(n)}(t) H(t)\right]_{d i a g} Q_{o} \tag{17}
\end{equation*}
$$

In the following we limit ourselves to the case of a one-dimensional model subspace, with $P_{o}=|i\rangle\langle i|$ and $Q_{o}=\sum_{k \neq i}|k\rangle\langle k|,|i\rangle$ being the initial state of the system. $X^{(n)}$ and $\delta X^{(n)}$ becomes vectors and $H_{e f f}(t)$ is a pure scalar function. Equation (16) becomes an ordinary differential equation whose rigorous solution for the component $\langle k| \delta X^{(n)}|i\rangle=\delta X_{k}^{(n)}$ can be expressed using exponential evolution operators,

$$
\begin{align*}
& \delta X_{k}^{(n)}(t)=\exp \left[\frac{1}{i \hbar} \int_{0}^{t}\left(\tilde{H}_{\text {diag }, k k}^{(n)}\left(t^{\prime}\right)-H_{e f f}^{(n)}\left(t^{\prime}\right)\right) d t^{\prime}\right]  \tag{18}\\
& \times \quad \int_{0}^{t}\left\{\Delta_{k}^{(n)}\left(t^{\prime}\right) \exp \left[-\frac{1}{i \hbar} \int_{0}^{t^{\prime}}\left(\tilde{H}_{\text {diag, } k k}^{(n)}\left(t^{\prime \prime}\right)-H_{e f f}^{(n)}\left(t^{\prime \prime}\right)\right) d t^{\prime \prime}\right]\right\} d t^{\prime} .
\end{align*}
$$

This solution is consistent with the initial value of the wave operator

$$
\begin{equation*}
\Omega(t=0)=P_{o}, \tag{19}
\end{equation*}
$$

hence $X^{(n)}(0)=0$. Equations (13), (17) and (18) are the central equations of this paper which are to be solved numerically.

### 2.2. Periodic basis set and constraints on the boundaries

Similar equations to those given above were derived in [19] by expanding the evolution operator using successive interaction pictures. However they were not applied numerically because of the difficulty of calculating the many time integrals which are similar to those present in (18). Only adiabatic limits could be calculated, with very poor results. Here we show that it is possible to design an efficient implementation by observing that the global approach represented by equations (7) and (18) is consistent with the use of a discrete Fourier basis set to span the whole propagation time interval, together with the associated periodic quadrature. These tools give effective ways to express the matrices $H(t) X^{(n)}(t)$ and $\partial X^{(n)}(t) / \partial t$ present in $\Delta^{(n)}(t)$ (equation (9)) and to calculate the integrals in (18), as will be explained in section 3. However the accurate use of Fast Fourier Transforms (FFT) is only possible if all the timedependent functions are made perodic and smooth.

The Hamiltonian $H(t)$ of a molecule submitted to a laser pulse is the sum of a time-independent part $H_{o}$ which represents the unperturbed molecule and of a perturbation $V(t)$ with $V(0)=V\left(T_{o}\right)=0$, where $T_{o}$ is the laser pulse duration. In equation (9) this corresponds to $X^{(n)}(t=0)=0$ and thus $\Delta^{(n)}(t=0)=0$. However at the end of the pulse duration, $X^{(n)}\left(t=T_{o}\right)$ can differ considerably from its initial value. To impose the continuity and the initial conditions of $X^{(n)}(t)$ one should use the CATM procedure [11, 12]. To this end, a time-dependent absorbing potential is introduced over an artificial time extension $\left[T_{o}, T\right]$. In the present case, this imaginary potential $V_{a b s}(t)$ destroys all the components of the wave operator (and the wavefunction), except the one corresponding to the initial state $\langle i \mid \Omega(t)\rangle$,

$$
\begin{equation*}
V_{a b s}(t)=-i V_{o p t}(t) \sum_{k \neq i}|k\rangle\langle k| \tag{20}
\end{equation*}
$$

where $V_{o p t}(t)$ is a real positive function localised on the interval $\left[T_{o}, T\right]$. This supplementary term produces the boundary value $X^{(n)}(T) \rightarrow 0$ and hence the periodicity of $\delta X^{(n)}$ and $\Delta^{(n)}$ (equation (9)):

$$
\begin{align*}
& \delta X^{(n)}(t=0)=\delta X^{(n)}(t=T)=0 \\
& \Delta^{(n)}(t=0)=\Delta^{(n)}(t=T)=0 \tag{21}
\end{align*}
$$

### 2.3. The problem of nested integrals over time

The $n^{\text {th }}$ order iteration requires solving the equations (13), (17) and (18) and the passage to the $(n+1)^{t h}$ order, via the incrementation of $X$ using $X^{(n+1)}=$ $X^{(n)}+\delta X^{(n)}$ and the construction of $\Delta^{(n+1)}$ (cf equation (9)). We want to use the same discretization at regularly distributed grid points to describe all the time dependencies. Suppose that the molecular basis set is composed of $N_{v}$ orthonormal states with $P_{o}=|i\rangle\langle i|$ and $Q_{o}=\sum_{k \neq i}|k\rangle\langle k|$ and that the discrete time-grid which spans the time-interval $[0, T]$ is composed of $N$ discrete time values $t_{j}=j T / N, j=0, \ldots, N-1$. The calculation of $\delta X^{(n)}(t)$ in (18) calls for the calculation of $N_{v}-1$ integrals $\int_{0}^{t_{j}}\left(\tilde{H}_{\text {diag,kk }}^{(n)}-H_{e f f}^{(n)}\right) d t^{\prime}$, where $k$ takes all the $N_{v}-1$ index values in the complementary space, followed by $N_{v}-1$ integrals $\int_{0}^{t_{j}} \Delta_{j}^{(n)}\left(t^{\prime}\right) e^{-\frac{1}{i \hbar} \int_{0}^{t^{\prime}}\left(\tilde{H}_{\text {diag,kk}}^{(n)}-H_{\text {eff }}^{(n)}\right) d t^{\prime \prime}} d t^{\prime}$. This is obviously a significant task, which must be undertaken at each interation order $(n)$. The calculation of each integral must be fast and very accurate. The selected procedure is explained in the next
section with a preliminary discussion about integrating an arbitrary complex function of time.

## 3. Discrete Fourier integration at the $n^{\text {th }}$ iteration order

In this section we explain how to compute efficiently a numerical definite integral of a time-dependent function described on regularly distributed sampling points. The function is not necessarily continuous at the boundaries of the grid. This preliminary discussion is important in facilitating the numerical implementation of equation (18). We require that the procedure should determine with the same accuracy, not only the integral at the final time value $t=T$ (which is easily obtained using the standard Fourier quadrature rule) but also its value for each of the $N$ intermediate discrete time values $t_{j}$.

### 3.1. Continuous integrals and numerical FFT-integrators

In the present subsection the Fourier transform $F(\nu)$ is defined as

$$
\begin{equation*}
F(\nu)=\int_{-\infty}^{+\infty} f(t) \exp (-i 2 \pi \nu t) d t \tag{22}
\end{equation*}
$$

Let $I(t)$ be the integral

$$
\begin{align*}
I(t) & =\int_{-\infty}^{t} f\left(t^{\prime}\right) d t^{\prime} \\
& =\int_{-\infty}^{+\infty} f\left(t^{\prime}\right) h\left(t-t^{\prime}\right) d t^{\prime} \tag{23}
\end{align*}
$$

where $h(t)$ is the Heaviside step function,

$$
h(t)= \begin{cases}0 & t<0 \\ 1 & t \geq 0\end{cases}
$$

The Fourier transform of $h(t)$ is the distribution

$$
\begin{equation*}
H(\nu)=\frac{1}{2} \delta(\nu)-P V\left(\frac{i}{2 \pi \nu}\right) \tag{24}
\end{equation*}
$$

PV denotes the Cauchy principal value which becomes relevant when the above expression is integrated. The convolution theorem gives

$$
\begin{align*}
I(t) & =f * h(t) \\
& =\int_{-\infty}^{+\infty} F(\nu) H(\nu) \exp (2 i \pi \nu) d \nu \tag{25}
\end{align*}
$$

Using $H(\nu)$ from equation (24) then gives

$$
\begin{equation*}
I(t)=\frac{1}{2} F(0)-\frac{i}{2 \pi} P V \int_{-\infty}^{+\infty} \frac{F(\nu) e^{2 i \pi \nu t}}{\nu} d \nu . \tag{26}
\end{equation*}
$$

In molecule-field interaction problems, the field term is frequently an oscillating function of time. Let us consider for a moment the special case of an oscillating function of the form $f(t)=g(t) \exp (i 2 \pi \tilde{\nu} t)$. Its Fourier transform is the translated function

$$
F(\nu)=G(\nu-\tilde{\nu})
$$

where $G(\nu)$ is the transform of $g(t)$. Using this $F(\nu)$ in (26) and using the variable change $\nu^{\prime}=\tilde{\nu}-\nu$, we obtain

$$
\begin{align*}
I(t) & =\int_{-\infty}^{t} g(t) \exp (i 2 \pi \tilde{\nu} t) d t=\frac{1}{2} G(-\tilde{\nu}) \\
& -\frac{i}{2 \pi} e^{i 2 \pi \tilde{\nu} t} P V \int_{-\infty}^{+\infty} \frac{G\left(-\nu^{\prime}\right) e^{-2 i \pi \nu^{\prime} t}}{\tilde{\nu}-\nu^{\prime}} d \nu^{\prime} \tag{27}
\end{align*}
$$

When the frequency $\tilde{\nu}$ is very high, we can neglect the Fourier coefficient $G(-\tilde{\nu})$ and approximate the denominator $\tilde{\nu}-\nu^{\prime}$ by $\tilde{\nu}$, giving the approximation

$$
\begin{equation*}
\int_{-\infty}^{t} g\left(t^{\prime}\right) \exp \left(2 i \pi \tilde{\nu} t^{\prime}\right) d t^{\prime} \underset{\tilde{\nu} \rightarrow \infty}{\sim} \exp [i(2 \pi \tilde{\nu} t-\pi / 2)] \frac{g(t)}{2 \pi \tilde{\nu}} \tag{28}
\end{equation*}
$$

We can now go back to the general case of equation (26) to derive a numerical FFT-integrator. The detailed derivation is given in Appendix A. This appendix explains how equation (26) can be implemented for periodic or non-periodic functions known as $N$ sample values distributed over the interval $[0, T]$. The integration algorithm needs only two FFT, with a computational cost scaling as $2 N \log N$ to obtain the $N$ values of $I\left(t_{j}\right)$, with $t_{j}=j T / N, j=0,1, \ldots, N-1$. Here we recall only the main result,

$$
\begin{equation*}
I\left(t_{j}\right)=F F T_{j}^{-1}(\mu . F F T(f))+\frac{j}{N} \times \frac{T}{\sqrt{N}} F F T_{k=0}(f) \tag{29}
\end{equation*}
$$

where $\mu \cdot F F T(f)$ is a component by component vector multiplication, with

$$
\mu_{\ell}=\left\{\begin{array}{l}
\frac{i}{2 \pi} \frac{T}{\ell} \quad \ell=1, \ldots \frac{N}{2}-1  \tag{30}\\
\frac{i}{2 \pi} \frac{T}{\ell-N} \quad \ell=\frac{N}{2}, \ldots N-1
\end{array}\right.
$$

In addition, the $\ell=0$ component of $(\mu . F F T(f))$ in (29) must be replaced by the following quantity:

$$
\begin{equation*}
a=-\sum_{\ell=1}^{N-1} \mu_{\ell} \cdot F F T_{\ell}(f) \tag{31}
\end{equation*}
$$

In the next subsection we perform some numerical tests using the algorithm derived in appendix Appendix A with rapidly oscillating functions.

### 3.2. Numerical test of the FFT-based integration algorithm

We study the arbitrary case of a function composed of six gaussians multiplied by complex exponential functions. The gaussians have various widths and are distributed on the interval $[0, T=180]$ to give the function

$$
\begin{equation*}
f(t)=\sum_{j=1}^{6} \exp \left[-a_{j}\left(t-t_{j}\right)^{2}\right] \exp \left(i 2 \pi x_{j} t / T\right) . \tag{32}
\end{equation*}
$$

The various parameters appearing in (32) are given in table 1. The $a_{j}$ are chosen so that the modulus of the function $f(t)$ is equal to or smaller than $10^{-20}$ at the two boundaries (see figure 1). In this example, the FFT procedure uses the function $\tilde{f}(t)$ defined by (A.6). This modified function is characterized by a spectrum with a central component $\ell=0$ equal to zero. This can be seen in figure 2 , which shows the modulus of the spectrum of $\tilde{f}(t)$. As the functions converge rapidly to zero at $t=0$ and at $t=T$, the continuity of the function $\tilde{f}(t)$ (assumed to be a periodic function

| Gaussian <br> index $j$ | $a_{j}$ | $t_{j}$ | $x_{j}$ |
| :--- | :--- | :--- | :--- |
| 1 | 6.790 | 27.000 | 12.000 |
| 2 | 3.819 | 36.000 | 135.600 |
| 3 | 1.018 | 90.000 | 1.750 |
| 4 | 1.591 | 108.000 | 154.700 |
| 5 | 2.118 | 135.000 | 3.250 |
| 6 | 3.310 | 144.000 | 18.150 |

Table 1. Parameters used in equation (32).


Figure 1. $|\Re e f(t)|$, see (32) and table 1.


Figure 2. Absolute value of the spectrum of $\tilde{f}(t)$ corresponding to (A.6) and (32). The function is given as a sample of $N=2048$ values equally distributed on the interval $[0, T]$.
by the FFT procedure) and also the continuity of the first two derivatives are ensured without the need to extend the time interval or to use an asymptotic polynomial (eq (A.10)). Figure 2 confirms that the functions are well represented by the Fourier expansion; the components omitted in the spectrum are smaller than $10^{-13}$.

We first compare the speed of convergence of two algorithms, the numerical FFTintegrator (this work) and the standard Simpson rule. As the exact values of the integrals $I_{j}=I\left(t_{j}\right)$ are not known, we perform two series of calculations by doubling the number $N$ of sampling points, comparing for each of the two integration procedures the result for $2 N$ points with that for $N$ points. The convergence factor $\mathrm{CF}_{N}$ is calculated by using the $N$ points (with even indices) that the $2 N$ grid has in common with the $N$ grid. At each of the N sampling points the absolute value of the difference between the two results for the integral is calculated and the convergence factor is defined as their maximum value, namely

$$
\begin{equation*}
\mathrm{CF}_{N}=\max _{j=0}^{N-1}\left(\left|I_{2 j}^{(2 N)}-I_{j}^{(N)}\right|\right) . \tag{33}
\end{equation*}
$$

The corresponding numerical results are shown in figure 3. As the number of sampling


Figure 3. Convergence factor (33) for the FFT algorithm (■) and the Simpson algorithm ( $\mathbf{\Delta}$ ).
points increases, the FFT convergence factor decreases rapidly by many orders of magnitude up to a critical $N$ value, $N_{F F T}^{c r i}=1024$. Beyond this point the convergence factor reaches a constant plateau with a precision close to machine precision $\left(10^{-14}\right)$. This is the usual behaviour expected from the Nyquist-Shannon theorem [26] when using Fourier basis expansions. The Simpson rule results show a strongly different behaviour. The convergence factor decreases regularly up to a minimum value for a critical value $N_{\text {Simp }}^{c r i}$, which is about 100 times larger than for the FFT case. Moreover the minimum convergence factor reached is about two orders of magnitude greater than for the FFT results. After this minimum value, the convergence factor increases again. In this new regime the roundoff errors become predominant. Since no exact result is available, we next compare the results obtained with the best precision for each method, to be sure that the two algorithms are consistent and converge to the same value. This corresponds to $N=1024$ for the FFT algorithm and to about
$N=150000$ for the Simpson algorithm. Figure 4 presents for each discrete time value the modulus of the difference between the two integrals calculated with the FFT algorithm and with the Simpson rule,

$$
\begin{equation*}
\left|I^{F F T}\left(t_{j}\right)-I^{\operatorname{Simp}}\left(t_{j}\right)\right| . \tag{34}
\end{equation*}
$$

This figure reveals that the difference on the whole interval $[0, T]$ is never larger than $10^{-13}$. Similar results, not presented here, are obtained for examples in which it was necessary to impose a continuity of the two first derivatives by using a $6^{\text {th }}$ order polynomial (see Appendix A.2). These applications prove that the FFT integrator


Figure 4. Difference between the FFT and the Simpson result: $\mid I^{F F T}\left(t_{j}\right)-$ $I^{\text {Simp }}\left(t_{j}\right) \mid$ (full line) and the modulus of the FFT result $\left|I^{F F T}\left(t_{j}\right)\right|$ (dashed line).
possesses the expected qualities. Figure 4 shows that the precision of the results is very high and virtually independent of the discrete instant $t_{j}$ taken as the upper bound. Moreover, figure 3 reveals that the number of sampling points necessary to give convergence for the FFT is 100 times less than that for the convergence of the Simpson algorithm. Finally the operations necessary for the FFT algorithm are reduced to the calculation of two FFT, the first being made in any case during the propagation scheme [27]. For the selected example the CPU time associated with the FFT procedure is about 20 times smaller than that associated with the Simpson rule. However, the degree of advantage of the FFT approach is closely related to the shape of the spectrum of $f(t)$. In the present cases as well as for the Schrödinger equation integration procedure treated in the next section, the spectrums are narrow and are thus well represented by a Fourier basis set.

## 4. An asymmetric double-well transition experiment

The illustrative example is that of a model diatomic molecule submitted to two laser pulses. Throughout this section we use arbitrary units with the convention $\hbar=1$. The various numerical parameters have been adjusted so that the relative orders of magnitude of the vibrational level spacing, the strength of the dipole couplings and the pulse duration produce realistic dynamics with significant transition probabilities.

Two potential curves $S=1,2$ are defined, which can refer to the two first electronic states of a one-dimensional vibrational Hamiltonian in the framework of the BornOppenheimer approximation. These two curves are shown in figure 5. The lower surface is a double well similar to those used in references [28, 29],

$$
\begin{equation*}
\epsilon_{1}(R)=-5 R^{2}+0.5 R^{3}+R^{4} . \tag{35}
\end{equation*}
$$

The upper surface is a single quartic well,

$$
\begin{equation*}
\epsilon_{2}(R)=0.2 R^{4} \tag{36}
\end{equation*}
$$

The $R$ coordinate is chosen as a relative position centred on the barrier of $\epsilon_{1}$. All the eigenstates are pure bound states and were calculated using a Fourier grid method on the radial coordinate $[30,31]$. They are shown in figure 5 . These model potentials are also similar to those used to describe, for instance, effective isomerization problems along a reaction coordinate [32]. The evolution of the vibrational evolution operator


Figure 5. The two potential energy curves and some of the first unperturbed vibrational eigenstates of the first potential surface (full lines) and of the second one (dashed lines). The potentials are quartic polynomials defined as: $\epsilon_{1}(R)=$ $-5 R^{2}+0.5 R^{3}+R^{4} ; \epsilon_{2}(R)=0.2 R^{4}$.
is driven by the following equation within the framework of the dipole approximation:

$$
i \frac{\partial}{\partial t} U(t, 0)=\left[T_{N}+\left(\begin{array}{cc}
\epsilon_{1} & -\vec{\mu}_{1,2} \cdot \vec{E}(t)  \tag{37}\\
-\vec{\mu}_{1,2} \cdot \vec{E}(t) & \epsilon_{2}
\end{array}\right)\right] U(t, 0)
$$

$T_{N}$ is the relative kinetic energy of the two atoms, i.e. $T_{N}=-\frac{1}{2 m} \frac{\partial^{2}}{\partial R^{2}}$ with an arbitrary mass equal to 10 . $\epsilon_{S=1,2}$ are the two energy surfaces. There is no rotation in the model and the dipole transition moment $\vec{\mu}$ is assumed to be constant along the radial axis $R$ and equal to 1 .

The laser field amplitude projected on the $\vec{\mu}$ direction is defined as the sum of two pulses with gaussian envelopes

$$
\begin{equation*}
E(t)=\sum_{j=1}^{2} E_{j} \cos \left(\omega_{j}\left(t-T_{j}\right)\right) \exp \left(-\left(\frac{t-T_{j}}{\tau_{j}}\right)^{2}\right) \tag{38}
\end{equation*}
$$

Vibrational states $(v=1, S=1)$ and $(v=6, S=1)$ have in common a strong overlap with state ( $v=7, S=2$ ) of the excited potential curve. We have therefore chosen the carrier frequencies as $\omega_{1}=E(v=7, S=2)-E(v=1, S=1) \simeq 9.98449$ and $\omega_{2}=E(v=7, S=2)-E(v=6, S=1) \simeq 4.77725$. This choice induces resonant transitions between the first and the sixth eigenstate of the lower surface: $\omega_{1}-\omega_{2}=E(v=6, S=1)-E(v=1, S=1)$. This will ensure strong transitions between the two asymmetric wells and a rather large depletion of the initial state. Calculations were made using $\tau_{1}=3.90$ and $\tau_{2}=4.50$ and two different electric field amplitudes and centres,

$$
\begin{align*}
& E_{1}=0.05, T_{1}=23.5 \\
& E_{2}=0.08, T_{2}=21.5 \tag{39}
\end{align*}
$$

in the first example and

$$
\begin{align*}
& E_{1}=0.09, T_{1}=22.5 \\
& E_{2}=0.05, T_{2}=21.5 \tag{40}
\end{align*}
$$

in the second example. The corresponding electric fields are represented in figure 6 . Those choices for the laser parameters place the system in a non optimal STIRAP configuration [33]. In both cases the initial state is the first eigenstate of the first surface ( $v=1, S=1$ ). In terms of the wave operator, the first global iteration starts with a guess chosen as $\Omega^{(0)}(t)=|v=1, S=1\rangle\langle v=1, S=1|$, i.e. $\quad X^{(0)}(t)=0$. A vibrational basis set made of the 30 first vibrational eigenvectors of each surface


Figure 6. The electric field $E(t)$ in the first example of (39) (left frame) and in the second example of (40) (right frame).
(i.e. a total molecular basis set of $N_{v}=60$ states) is sufficient to obtain converged results. This molecular basis set is small but here the principal aim is to test the time propagation algorithm. Here we need $N_{t}=4096$ Fourier states $|p\rangle$ with $\langle t \mid p\rangle=\exp (i 2 \pi p t / T)$ (or equivalently $N_{t}=4096$ time grid points) to describe the time evolution correctly. In what follows the vibrational states are numbered from $v=1$ to $v=30$ for the first surface and from $v=31$ to $v=60$ for the second surface. The total basis set describing the extended Hilbert space is then composed of $N=N_{v} \times N_{t}=245760$ states. In this framework, the dynamics is calculated by using the iterative scheme explained in section 2 (equations (5), (13), (17) and (18) with $\Omega^{(n+1)}=P_{o}+X^{(n+1)}$ and $X^{(n+1)}=X^{(n)}+\delta X^{(n)}$.) The integrals appearing in (18) are calculated using the numerical integrator presented in section 3 and Appendix A. This scheme is used to calculate all the components of the off-diagonal part of the wave
operator for each of the $N_{t}$ discrete $t_{k}$ values: $X_{v}^{(n)}\left(t_{k}\right), v=1, \ldots, N_{v}, k=1, \ldots, N_{t}$, at each iteration order $(n)$.


Figure 7. Convergence factor (41) with the parameters defined in (39) and those defined in 40 ( $\square$ ) with respect to the iteration number.

Our main concern is the speed of convergence of the global iterative scheme. We define the following convergence factor:

$$
\begin{equation*}
F=\frac{\sum_{v=1}^{N_{v}} \sum_{k=1}^{N_{t}}\left|\delta X_{v}^{(n)}\left(t_{k}\right)\right|^{2}}{\sum_{v=1}^{N_{v}} \sum_{k=1}^{N_{t}}\left|X_{v}^{(n+1)}\left(t_{k}\right)\right|^{2}} \tag{41}
\end{equation*}
$$

with $\delta X^{(n)}$ defined in (18). Figure 7 shows that the two calculations converge. A plateau of about $10^{-17}$ for the first example and of about $10^{-16}$ for the second one is reached at the $22^{t h}$ iteration. Between the $10^{t h}$ and the $22^{t h}$ iteration, the accuracy is improved by one order of magnitude at each iteration. The convergence factor of equation (41) adds the residues $\delta X^{(n)}(t)$ over the whole propagation interval. This means that the plateau seen in figure 7 indicate a very high accuracy at each time grid point. It is essential to note that, by contrast with a continuous propagation scheme, in which errors are accumulated at each time step, the global scheme does not accumulate the errors between two iteration orders. Nevertheless our iterative procedure possesses, as does every iterative treatment, a finite radius of convergence. The small increase of the convergence factor observed in figure 7 during the first four iterations for the second example is the consequence of a large Fubini-Study distance between $P_{o}$ and $P$ in this case (the survival probability at the end of the interaction $\left|\left\langle v=1 \mid \Psi\left(t_{\text {final }}\right)\right\rangle\right|^{2}$ is 0.2129 in the first example but only 0.02864 in the second one). This increases the distance between $P_{o}$ and $P\left(t_{\text {final }}\right)$ in this last case. This can be seen in figure 8 which illustrates the progressive separation between the active subspace $S_{o}$ and the dynamic subspace $S(t)$. In the non-degenerate case, the Fubini-Study distance is simply

$$
\begin{equation*}
\operatorname{dist}_{F S}\left(P_{o}, P(t)\right)=\arccos \left(\left\|\left\langle v_{i} \mid \Psi(t)\right\rangle\right\|\right)=\arccos \left(\frac{1}{\|\Omega(t)\|}\right) . \tag{42}
\end{equation*}
$$

The right-hand side is obtained using the normalization properties of the wave operator. The final distance for the second example of (40) is clearly larger. The


Figure 8. Evolution of the Fubini-Study distance $\operatorname{dist}_{F S}\left(P_{o}, P(t)\right)$ during the pulse defined in (39) (solid line) and (40) (dashed line). The distance has been computed using the wave operator at iteration 22.
physical reason explaining this strong difference is that example (40) is closer to an efficient STIRAP situation than example (39) (but still non-optimal, the final transition probability to state $|v=6\rangle$ being here only 0.22 , because of many nonadiabatic transitions to other vibrational states).

The global character of the integration is evident in figures 9 and 10 which present the transition moduli to the different vibrational states at the final time $\left|\left\langle v \mid \Psi^{(n)}\left(t_{\text {final }}\right)\right\rangle\right|$ at four different iteration orders. It is interesting to note that in figure 9 , an approximate vibrational distribution close to the exact one is obtained as early as at the fourth iteration. The small convergence difficulties during the first iteration orders in the second example are evident in figure 10. The $4^{t h}$ order shows some transition amplitude larger than 50 , very far from the final results. In this case more iterations are needed to achieve a stabilized distribution. Probabilities can thus exceed one in the global scheme if the results are observed at an early intermediate order, before the completion of the iterative process, such as in Fig. 10. There is no fixed norm conservation in the structure of the algorithm but the norm conservation is recovered after convergence. Figures 11 and 12 also illustrate the global character of the time-dependent propagation. In these figures we present the evolution of the transition amplitude to state $|v=37\rangle$. This particular state is strongly coupled with the initial state $|v=1\rangle$. Figure 11 shows that each iteration order produces a projected amplitude which converges towards the exact solution over the whole interaction time interval. At the fourth iteration, the difference $\left|\left\langle v=37 \mid \Psi^{(n=22)}(t)-\Psi^{(n)}(t)\right\rangle\right|$ is already lower than $10^{-1}$. At the $9^{t h}$ order, the same error becomes lower than $10^{-3}$. Figure 12 reveals the difficulties due to the large Fubini-Study distance between $P_{o}$ and $P\left(t_{\text {final }}\right)$ (here the final survival probability is $\left.\mid\left.\langle v=1| \Psi\left(t_{\text {final }}\right\rangle\right|^{2}=0.02864\right)$. As a consequence, the amplitude of the projected wave function onto the state $v=37$ at the $4^{t h}$ iteration order is about 10 times larger than the exact one for $t>30$. Surprisingly, the iterative procedure converges again between the $4^{\text {th }}$ and the $9^{t h}$ order. At the $9^{\text {th }}$ order the error is smaller than $10^{-2}$


Figure 9. Final transition amplitudes to the different vibrational states $\left|\left\langle v \mid \Psi^{(n)}\left(t_{\text {final }}\right)\right\rangle\right|$ after the pulse defined in (39), obtained from the wave operator at successive iteration orders. $+:(n=2), \times:(n=4), *:(n=9), \bullet:(n=22)$.


Figure 10. Final transition amplitudes to the different vibrational states $\left|\left\langle v \mid \Psi^{(n)}\left(t_{\text {final }}\right)\right\rangle\right|$ after the pulse defined in (40), at successive iteration orders. $+:(n=2), \times:(n=4), *:(n=9), \bullet:(n=22)$.
everywhere.
We think that this efficiency is essentially due to the use of a time-dependent effective Hamiltonian in the basic equations of the integration scheme ((13), (17) and (18)), especially when it is compared with previous iterative schemes. Figure 13 reveals large oscillation of the effective Hamiltonian defined in (5). The imaginary part imposes the unitarity of the evolution in the wave operator formulation. A contribution to the evolution operator of the form

$$
\exp \left(\frac{1}{i \hbar} \int_{0}^{t_{f i n a l}} \Im m\left(H_{e f f}\right) d t^{\prime}\right)
$$



Figure 11. The dashed line represents the modulus of the projected wave function $\left|\left\langle v=37 \mid \Psi^{(n)}(t)\right\rangle\right|$ as a function of time at the end of the iterative process ( $n=22$ ) for the first example (parameters given in (39)). The full lines are the defects at intermediate iteration orders: $\left|\left\langle v=37 \mid \Psi^{(n=22)}(t)-\Psi^{(n)}(t)\right\rangle\right|$ for ( $n=4$ ) and ( $n=9$ ).


Figure 12. The dashed line represents the modulus of the projected wave function $\left|\left\langle v=37 \mid \Psi^{(n)}(t)\right\rangle\right|$ as a function of time at the end of the iterative process $(n=22)$ for the second example (parameters given in (40)). The thin continuous lines are the defects at intermediate iteration orders : $\mid\langle v=$ $37\left|\Psi^{(n=22)}(t)-\Psi^{(n)}(t)\right\rangle \mid$ for $(n=4)$ and $(n=9)$. The deep black line shows the survival amplitude $\left|\left\langle v=1 \mid \Psi^{(n=22)}(t)\right\rangle\right|$.


Figure 13. Evolution of the real part (continuous line) and imaginary part (dashed line) of the effective energy, using the field parameters (40).
is present and gives the correct survival amplitude. The magnitude of these oscillations ( $\approx 4$ for the real part) is even much larger than the distance between the first closest vibrational eigenvalues of each surface (for example $E(v=2)-E(v=1)=1.451$ ). Other iterative methods in which these time variations are neglected completely fail to converge on this same numerical example. For example, the adiabatic approximation would be justified if in equation (18) the amplitude $\left.\left|\langle v| \tilde{H}_{\text {diag }}^{(n)}(t)\right| v\right\rangle-\left\langle v_{i}\right| H_{\text {eff }}^{(n)}(t)\left|v_{i}\right\rangle \mid$ is large at each instant, $\left|v_{i}\right\rangle$ being the initial vibrational state. In such a case the introduction of the approximation (28) into equation (18) leads to:

$$
\begin{equation*}
\delta X^{(n)}(t)=\frac{i \hbar \Delta^{(n)}(t)}{\tilde{H}_{\text {diag }}^{(n)}(t)-H_{e f f}^{(n)}(t)} \tag{43}
\end{equation*}
$$

Equation (43) is very similar to the RDWA iterative method which has already been used to compute time-dependent wavepacket propagations [15, 14]. However it cannot be used in the present case. It is evident that the strong time-dependency of $H_{e f f}(t)$ (see figure 13) and the very small survival probability at $t_{\text {final }}$ are inconsistent with the use of any adiabatic approximation. Actually the use of (43) in the integration scheme produces a very fast divergence of the iterative series caused by the crossings of $\tilde{H}_{\text {diag }}^{(n)}$ and $H_{\text {eff }}^{(n)}$. Another approximate (but more sophisticated) integration scheme was proposed in refs [17] and [15]. By using a discrete Fourier basis set: $|p\rangle$ with $\langle t \mid p\rangle=\exp (i 2 \pi p t / T)$ derived from the discrete time grid basis set by a FFT transformation, a generalised RDWA approximation leads to:

$$
\begin{align*}
& \left\langle v^{\prime}, p^{\prime}\right| X^{(n+1)}\left|v_{i}, p=0\right\rangle \\
& =\frac{\left\langle v^{\prime}, p^{\prime}\right|\left[\left(H_{F}-\tilde{H}^{(n)}\right) X^{(n)}+H\right]\left|v_{i}, p=0\right\rangle}{\left\langle v_{i}, p=0\right| H_{e f f}^{(n)}\left|v_{i}, p=0\right\rangle-\left\langle v^{\prime}, p^{\prime}\right| \tilde{H}^{(n)}\left|v^{\prime}, p^{\prime}\right\rangle} \tag{44}
\end{align*}
$$

Introducing (44) into the iterative scheme also produces a divergence, albeit slower than the one produced by (43). Undoubtedly this defect comes from the fact that equation (44) only uses an averaged $H_{\text {eff }}$, represented by the first Brillouin component $\langle p=0| H_{e f f}|p=0\rangle$, rather than the exact and strongly time-dependent expression.

Finally it is crucial to verify that the global scheme gives the correct results by a comparison with some continuous wave-packet propagation algorithms. We first compare our results with those obtained for the same system by using a split operator [7] procedure by separating the role of the diagonal unperturbed Hamiltonian and the role of the laser-molecule coupling. This last contribution is not diagonal and its role is approximated by using a Second Order Differencing (SOD) scheme. We also calculate the same dynamics using the short iterative Lanczos (SIL) method [8]. To be significant, the same vibrational basis set of $N_{v}=60$ states is used in the continuous propagations (for the SIL propagations, the size of the Lanczos subspace was set up to 10). A convergence factor is defined as


Figure 14. Convergence factors $\mathcal{F}_{C}$ and $\mathcal{F}_{G}$ defined in equations (45) and (46) for the global method (circles), the short iterative Lanczos scheme (triangles) and the split-operator algorithm (squares) using the field parameters of equation (39) ( $\square, \triangle$ and $\circ$ ) and of equation (40) ( $\square, \Delta$ and $\bullet$ ), as a function of the number of time steps.

$$
\begin{equation*}
\mathcal{F}_{C}\left(N_{t}\right)=\sum_{i=1}^{N_{v}}\left|\Omega^{C, N_{t}}\left(i, t_{\text {final }}\right)-\Omega^{G, 4096}\left(i, t_{\text {final }}\right)\right|^{2} \tag{45}
\end{equation*}
$$

$\Omega^{G, 4096}$ is the wave operator calculated at the final time using the global propagator with 4096 time grid points after 22 iterations and $\Omega^{C}$ is the wave operator reconstructed by using equation (4) after a complete propagation with the continuous propapagators with $N$ time steps. To show the convergence of the global method itself we have also computed a similar convergence factor between global results,

$$
\begin{equation*}
\mathcal{F}_{G}\left(N_{t}\right)=\sum_{i=1}^{N_{v}}\left|\Omega^{G, N_{t}}\left(i, t_{\text {final }}\right)-\Omega^{G, 8192}\left(i, t_{\text {final }}\right)\right|^{2} \tag{46}
\end{equation*}
$$

Figure 14 shows the evolution of $\mathcal{F}_{C}$ and $\mathcal{F}_{G}$ with respect to the number of discrete time steps. This figure reveals that the three procedures converge to the same final values (same amplitudes and same phases). Figure 14 also confirms that the number of time steps needed to converge to a given accuracy is much smaller with the global scheme than with the continuous schemes. The convergence is faster for the first pulse
than for the second one. The number of time-steps of length $d t$ for the continuous integrator is quite large especially if a high accuracy is needed. The CPU time needed to compute $\delta X^{(n)}$ (equation (18)) at a given iteration order is of course much larger than that required to construct one step in the continuous algorithms. Nevertheless the convergence plateau in figure 7 are reached after only 22 iterations. With the field parameters of equation (39), passing from iteration 22 to iteration 23 indicates that the transition amplitudes are converged with 7 stable and significant digits. More detailed informations about the CPU time are given in table 2 with varying required accuracies. This table shows the ratio between the CPU times required to reach a given threshold of convergence using the global or the continuous strategies, using the convergence factors $\mathcal{F}_{G}$ and $\mathcal{F}_{C}$ defined in (45) and (46). The results of the global method have been observed after 22 iterations. The continuous algorithms are faster

| Convergence factor $<$ | $10^{-6}$ | $10^{-8}$ | $10^{-10}$ | $10^{-12}$ | $10^{-14}$ |
| :--- | :--- | :--- | :--- | :--- | :--- |
| Stable digits $\simeq$ | 3 | 4 | 5 | 6 | 7 |
| $T_{G}^{C P U} / T_{S P L U}^{C P U}$ | 3.83 | 3.11 | 0.775 | 0.757 | 0.211 |
| $T_{G}^{C P U} / T_{\text {Split-Op. }}^{C P}$ | 15.3 | 6.52 | 3.31 | 1.62 | 0.912 |

Table 2. Approximative ratios of the CPU times required to reach a given convergence threshold between the global (G) and the continuous methods (SIL or split-operator), for example of equation (39). The corresponding number of significant digits on the largest transition probabilities are also given.
if a low accuracy is sufficient. The SIL algorithm and the global method then give comparable CPU times for intermediate accuracies, while the split operator stays a little faster. The global method becomes slightly faster if a high accuracy is desired.

## 5. Conclusion

The theoretical analysis and the simple illustrative examples presented in this paper demonstrate that a short global iterative scheme can solve the time-dependent Schrödinger equation with an explicitly time-dependent Hamiltonian operator. This global method is competitive with a continuous step by step algorithm. The following points should be highlighted:
i) A global integrator does not accumulate discretization errors along the time interval and consequently can generate high precision results. The numerical experiments have shown that each new iteration order increases the precision by one order of magnitude. The structure of the algorithm also allows for a rigourous estimation of the error associated with each iteration order by following the quantity $H_{F} \Omega-\Omega H \Omega$, the calculation being stopped when the expected accuracy is reached.
ii) The use of an FFT procedure to calculate the numerous time integrals appears as a very efficient choice which is compatible with producing a highly accurate solution to the Schrödinger equation. Nevertheless it is essential that the Fourier spectrum of the perturbation should be well reproduced by the Fourier basis set and it is certainly more difficult to solve problems with broader spectra.
iii) The global iterative scheme does not introduce limitations concerning the nature or the structure of the Hamiltonian which drives the dynamics. Future applications could usefully be aimed at checking how well it can reproduce dissipative dynamics.
iv) The inability of previous global iterative techniques to determine the timedependent wave-operator for similar problems is partly due to a wrong estimation of the effective Hamiltonian acting in the selected model space. In the selected examples $H_{e f f}(t)$ exhibits large oscillations which must correctly be taken into account.
v) In the present applications the starting input for the wave-operator is simply $X^{(n=1)}=0$. The iterative nature of the method makes possible a rapid generation of new solutions, after a slight modification of the Hamiltonian parameters, by simply starting with the wave-operator obtained from the previous solution. Many applications are possible, among which we can mention the study of laser control processes using exceptional points of parametric Hamiltonians [34, 35, 36, 37] or of vibrational cooling using zero-width resonances [38].

The algorithm proposed in this paper appears to be useful for investigating new formulations of quantum control in molecular physics or for treating complicated nearadiabatic evolutions, for which the use of an effective Hamiltonian can efficiently reduce the dimension of the Hilbert space [39]. The global algorithm add the time as a new quantum coordinate, which can be a disadvantage to compute the dynamics of high dimensional problems over long times. On the other hand the method is not dependent on the structure of the Hamiltonian as it can be the case for the split operator algorithm or Magnus expansions. The wave operator approach is in principle transposable to dynamical processes like dissociation, ionization, dynamics involving degenerate eigenvalues or non-adiabatic interactions. For dissipative processes, a special difficulty can appear in situations for which the dynamics escapes too far from the selected model subspace. In such cases, the algorithm can diverge. The best option is to generalise the iterative algorithm of section 2 to handle a degenerate active space, by incorporating all the states which are strongly coupled into the model space. Study of this generalisation is in progress, as the adaptation of the algorithm to the calculation of Floquet eigenstates.

## Acknowledgments

We thank John P. Killingbeck for fruitful discussions and for a careful reading of the English manuscript. Simulations have been executed on computers of the Utinam Institute of the Université de Franche-Comté, supported by the Région de FrancheComté and Institut des Sciences de l'Univers (INSU).

## Appendix A. Numerical FFT integration

## Appendix A.1. Integration of periodic functions

Consider equation (26),

$$
\begin{equation*}
I(t)=\frac{1}{2} F(0)-\frac{i}{2 \pi} P V \int_{-\infty}^{+\infty} \frac{F(\nu) e^{2 i \pi \nu t}}{\nu} d \nu \tag{A.1}
\end{equation*}
$$

where $F(\nu)$ is the Fourier transform of $f(t)$. If the function $f(t)$ is assumed to be nonzero only over a finite interval $[0, T]$, then a discrete finite time-grid can be introduced to span this time-interval and also the corresponding finite frequency representation. For even $N$ we have:

$$
t_{j} \quad=\frac{j T}{N}, j=0, \ldots, N-1
$$

$$
\begin{align*}
& \nu_{j}=\frac{j}{T}, j=0, \ldots, \frac{N}{2}-1 \\
& \nu_{N / 2}=-\frac{N}{2 T} \\
& \nu_{j}=-\nu_{N-j}, j=\frac{N}{2}+1, \ldots, N-1 \tag{A.2}
\end{align*}
$$

If the time $t$ and frequencies $\nu$ are selected as one of the points of the above discrete grid, a discrete version of the integral can be computed. The continuous Fourier transform is approximated by a discrete Fast-Fourier Transform (FFT):

$$
\begin{align*}
& \int_{-\infty}^{+\infty} f(t) e^{-2 i \pi \nu_{k} t} d t= \\
& \int_{0}^{T} f(t) e^{-2 i \pi \nu_{k} t} d t \approx \frac{T}{\sqrt{N}} F F T_{k}(f) \tag{A.3}
\end{align*}
$$

where

$$
\begin{equation*}
F F T_{k}(f)=\frac{1}{\sqrt{N}} \sum_{j=0}^{N-1} f_{j} e^{-i 2 \pi j k / N} \tag{A.4}
\end{equation*}
$$

with $f_{j}=f\left(t_{j}\right)$. The integral in equation (26) is then apparently approximated by the sum

$$
\begin{equation*}
I\left(t_{j}\right)=\frac{1}{2} \frac{T}{\sqrt{N}} F F T_{k=0}(f)-\frac{i}{2 \pi} \frac{1}{\sqrt{N}} \sum_{\ell=0}^{N-1} \frac{F F T_{\ell}(f)}{\nu_{\ell}} e^{2 i \pi \ell j / N} \tag{A.5}
\end{equation*}
$$

However, the passage from the continuous expression (26) to the FFT equivalent (equation (A.5)) has not been carried out with sufficient rigour, since there is a division by 0 for $\ell=0$ in (A.5). We must adjust the term $\ell=0$ before doing the inverse transform. The use of a Cauchy principal value and the presence of a singularity in $(26)$ at $\nu=0$ if $F(\nu=0) \neq 0$ are inconsistent with the use of a Fourier series expansion. The equations should thus be modified to take into account this discrepancy. However, we can preserve the feature that only two Fourier transforms are required to obtain the integral.

The calculation of the $N$ integrals $I\left(t_{j}\right), j=0, \ldots, N-1$ is made as follows. The Fourier transform $F F T(f)$ is first calculated and then a new function $\tilde{f}(t)$ is defined:

$$
\begin{equation*}
\tilde{f}(t)=f(t)-\frac{1}{\sqrt{N}} F F T_{k=0}(f) \tag{A.6}
\end{equation*}
$$

Both $f$ and $\tilde{f}$ have the same discrete spectrum, except for the $\ell=0$ component, which for $\tilde{f}$ is equal to zero, so that the integral of $\tilde{f}(t)$ over the interval $[0, T]$ is equal to zero. This is equivalent to using a translation of the ordinate axis. We work temporarily with $\tilde{f}$ instead of $f$. We assume that the spectrum $\tilde{F}(\nu)$ (equal to zero for $\nu=0$ ) has a Taylor expansion near $\nu=0$ :

$$
\tilde{F}(\nu)=a \nu+b \nu^{2}+\ldots
$$

We are looking for the finite coefficient $a=\lim _{\nu \rightarrow 0} \frac{\tilde{F}(\nu)}{\nu}$, so as to apply equation (A.5) at $\ell=0$ in a consistent way. The divergence produced by the term $\nu_{\ell} \propto \frac{1}{\ell}$ for $\ell=0$ in (A.5) has disappeared but the value of $a$ is still unknown. One can guess it by
requiring that $I\left(t_{0}=0\right)=\int_{0}^{0} \tilde{f}(t) d t=0$. Equation (A.5) in this case leads to the value

$$
\begin{equation*}
a=-\sum_{\ell=1}^{N-1} \mu_{\ell} \cdot F F T_{\ell}(f) \tag{A.7}
\end{equation*}
$$

where $\mu \cdot F F T(f)$ denotes a component by component vector multiplication, with

$$
\mu_{\ell}=\left\{\begin{array}{l}
\frac{i}{2 \pi} \frac{T}{\ell} \quad \ell=1, \ldots \frac{N}{2}-1  \tag{A.8}\\
\frac{i}{2 \pi} \frac{T}{\ell-N} \quad \ell=\frac{N}{2}, \ldots N-1
\end{array}\right.
$$

This value of $a$ replaces the $\ell=0$ component of ( $\mu \cdot F F T(f)$ ) in equation (A.9) below. So far we have ignored the zero frequency component of the function by working with $\tilde{f}$ instead of $f$. The final result should then be corrected by a linear correction function, giving

$$
\begin{equation*}
I\left(t_{j}\right)=F F T_{j}^{-1}(\mu . F F T(f))+\frac{j}{N} \times \frac{T}{\sqrt{N}} F F T_{k=0}(f) \tag{A.9}
\end{equation*}
$$

The first term on the right hand side represents the integral of $\tilde{f}$ at the discrete time values $t_{j}$ and the second term restores the correct value of $\int_{0}^{t_{j}} f(t) d t$. Equations (A.9), (A.8) and (A.7) are the basic equations needed to calculate the $N$ integrals $I\left(t_{j}\right)$ using two FFT calls (with a computational cost scaling as $2 N \log N$ ).

The above equations have close similarities with those of the FFT-based integration algorithms proposed in ref. [20, 21] in the framework of image resampling. The derivation starting from (26) emphasizes the difficulties introduced by the Cauchy principal value term for small frequencies. Here the principal value integral is approximated by first cancelling $\tilde{F}(\nu=0)$ and guessing its first derivative at $\nu=0$ in a consistent and unambiguous way. Our expressions also work well for functions with integrals over the interval $[0, T]$ which are not necessarily equal to zero.

## Appendix A.2. Integration of non-periodic functions

In the algorithm developed in section 2 the smoothness of the wave operator and the wavefunction at the time boundary $t=T$ is ensured by the presence of a timedependent absorbing potential (cf. section 2.2). However the intermediate matrix elements of $H_{e f f}^{(n)}(t)$ and $\tilde{H}_{\text {diag, } k k}^{(n)}(t)$ to be integrated are not necessarily T-periodic. FFT-based differentiation and integration algorithms are very accurate but suffer from boundary effects (Gibb's phenomenon). The present algorithm will give accurate results only with periodic continuous functions, ideally those with time derivatives which are themselves also continuous functions. There are different ways of adapting a Fourier-based algorithm to deal with non-periodic signals. For example, extension methods have been developed to solve partial differential equations on a complex domain [22, 23], to represent surfaces of complex bodies [24] and to treat imaging problems [20].

Here we use the alternative procedure of Hermite interpolants to ensure the continuity of the function $f(t)$ and of the first derivatives [25]. Let $f(t)$ be a complex continuous $L^{2}$ function defined on a large time interval $\left[0, T_{0}\right.$ ] , and represented by a sample of $N_{0}$ equally distributed values $f\left(t_{j}\right), \quad t_{j}=j T_{0} / N_{0}, j=0, \ldots, N_{0}-1$ on this interval. The procedure described in (A.6) has been applied to $f(t)$ so that $F F T_{k=0}(f)=0$. As the function $f(t)$ is assumed to be continuous with continuous first and second derivatives inside the interval $\left[0, T_{0}\right]$, the principal problem is to ensure
continuity at the boundaries. Let $f(0), f^{\prime}(0), f^{\prime \prime}(0)$ and $f\left(T_{0}\right), f^{\prime}\left(T_{0}\right), f^{\prime \prime}\left(T_{0}\right)$ be the values of the function and of the first two derivatives at the two boundaries. It will be impossible to obtain accurate results if these values do not match. One can extend the time interval by adding a narrow extra interval $\left[T_{0}, T\right]$ spanned by an extension of the grid $t_{j}=j T_{0} / N_{0}, j=N_{0}, \ldots, N-1$ and then constructing on this new interval a $6^{t h}$ order polynomial:

$$
\begin{equation*}
f(t)=\sum_{j=0}^{6} a_{j}\left(t-T_{0}\right)^{j} . \tag{A.10}
\end{equation*}
$$

Seven conditions are imposed on the polynomial. The first three are related to the continuity of $f(t)$ at the point $t_{j}=T_{0}$ and give the first three coefficients $a_{j}$ :

$$
\begin{align*}
a_{o} & =f\left(T_{0}\right), \\
a_{1} & =f^{\prime}\left(T_{0}\right),  \tag{A.11}\\
a_{2} & =f^{\prime \prime}\left(T_{0}\right) / 2
\end{align*}
$$

The three values on the right hand sides of (A.11) are known exactly if the function $f(t)$ has an analytical expression. Alternatively, the derivatives can be approximated by a FFT algorithm if $f(t)$ is known at a sample of $N_{0}$ values. The last four coefficients are related to the continuity between $t=T$ and $t=0$ of $f(t)$ and to the condition $\int_{0}^{T} f(t) d t=0$ :

$$
\begin{align*}
& f(T)=f(0) \\
& f^{\prime}(T)=f^{\prime}(0) \\
& f^{\prime \prime}(T)=f^{\prime \prime}(0)  \tag{A.12}\\
& \int_{T_{0}}^{T} f(t) d t=0
\end{align*}
$$

With $\Delta=T-T_{0}$, the last four unknowns are found by solving the small linear system:

$$
\begin{align*}
& \left(\begin{array}{cccc}
\Delta^{3} & \Delta^{4} & \Delta^{5} & \Delta^{6} \\
3 \Delta^{2} & 4 \Delta^{3} & 5 \Delta^{4} & 6 \Delta^{5} \\
6 \Delta & 12 \Delta^{2} & 20 \Delta^{3} & 30 \Delta^{4} \\
\frac{\Delta^{4}}{4} & \frac{\Delta^{5}}{5} & \frac{\Delta^{6}}{6} & \frac{\Delta^{7}}{7}
\end{array}\right)\left(\begin{array}{l}
a_{3} \\
a_{4} \\
a_{5} \\
a_{6}
\end{array}\right) \\
& =\left(\begin{array}{c}
f(0)-a_{0}-a_{1} \Delta-a_{2} \Delta^{2} \\
f^{\prime}(0)-a_{1}-2 a_{2} \Delta \\
f^{\prime \prime}(0)-2 a_{2} \\
-a_{0} \Delta-a_{1} \frac{\Delta^{2}}{2}-a_{2} \frac{\Delta^{3}}{3}
\end{array}\right) \tag{A.13}
\end{align*}
$$

Note that the last (integral) condition (A.12) is not absolutely necessary and can be relaxed by first defining the above polynomial and then applying the zero-integral condition on the whole interval using equation (A.6).

In summary, the FFT-based integration algorithm is implemented using (A.7)(A.9), completed by a preliminary treatment on the additional interval $\left[T_{0}, T\right]$ using (A.10)-(A.13).

## References

[1] Sun Y and Mowrey R C 1987 J. Chem. Phys. 86 2087; Sun Y Judson R S and Kouri D J 1989 J. Chem. Phys. 90, 241
[2] Neuhauser D and Baer M 1990 J. Chem. Phys. 923419
[3] Atabek O Lefebvre R and Nguyen-Dang T-T 2010 Adv. Quantum Chem. 60 51-104
[4] Meyer H-D Manthe U and Cederbaum L S 1990 Chem. Phys. Lett. 165 73; Meyer H-D Gatti F and Worth G A 2009 Multidimensional Quantum Dynamics: MCTDH Theory and Applications (Wiley-VCH, Weinheim)
[5] Leforestier C et al 1991 J. of Comp. Phys. 9459
[6] Askar A and Cakmak A S 1978 J. Chem. Phys. 682794
[7] Feit M D Fleck J A and Steiger A 1982 J. Comp. Phys. 47412
[8] Park T J and Light J C 1986 J. Chem. Phys. 855870
[9] Sanz-Serna J and Portillo A 1995 J. Chem.Phys. 1042349
[10] Kormann K, Holmgren S and Karlsson H O 2008 J. Chem.Phys. 128184101
[11] Jolicard G Viennot D and Killingbeck J P 2004 J. Phys. Chem. A 1088580
[12] Leclerc A Guérin S Jolicard G and Killingbeck J P 2011 Phys. Rev. A 83032113
[13] Leclerc A Jolicard G and Killingbeck J P 2011 J. Chem. Phys. 134194111
[14] Leclerc A Jolicard G Viennot D and Killingbeck J P 2012 J. Chem. Phys. 136014106
[15] Jolicard G and Killingbeck J P 2003 J. Phys. A 36 R411-R473
[16] Viennot D 2014 J. Phys. A. 47065302
[17] Jolicard G Viennot D Killingbeck J P and Zucconi J M 2004 Phys. Rev. E 70046703
[18] Killingbeck J P 1991 Microcomputer Algorithms Action from Algebra (Adam Hilger, Bristol)
[19] Jolicard G and Austin E 1991 Chem. Phys. Lett. 180503
[20] Yaroslavsky L P Moreno A and Campos J 2005 Optics Express 132892
[21] Yaroslavsky L P 2007 Fast discrete sinc-interpolation: a gold standard for image resampling in Advances in Signal Transform: Theory and Applications 7 337-405
[22] Elghaoui M and Pasquetti R 1996 J. Comp. Phys. 125464
[23] Averbuch A Vozovoi L and Israeli M 1997 Num. Alg. 15287
[24] Bruno O P Han Y and Pohlman M N 2007 J. Comp. Phys. 2271094
[25] Boyd J P 2002 J. Comp. Phys. 178118
[26] Shannon C E 1949 Proc. IRE 3710
[27] Kosloff R 1996 Quantum Molecular Dynamics on Grids chapter V Dynamics of molecules and chemical reactions eds R E Wyatt and J Z H Zhang (Marcel Dekker Inc)
[28] Bavli R and Metiu H 1993 Phys. Rev. A 473299
[29] Wadehra A Vikas and Deb B M 2003 J. Chem. Phys. 1196620
[30] Marston C C and Balint-Kurti G G 1989 J. Chem. Phys. 913571
[31] Kosloff D and Kosloff R 1983 J. Comp. Phys. 5235
[32] Chenel A Dive G Meier C and Desouter-Lecomte M 2012 J. Phys. Chem. A 11611273
[33] Vitanov N V Halfmann T Shore B W Bergmann K 2001 Annu. Rev. Phys. Chem. 52763
[34] Lefebvre R Atabek O Sindelka M and Moiseyev N 2009 Phys. Rev. Lett. 103123003
[35] Uzdin R Mailybaev A and Moiseyev N 2011 J. Phys. A: Math. Gen. 44435302
[36] Leclerc A Jolicard G and Killingbeck J P 2013 J. Phys. B: At. Mol. Opt. Phys. 46145503
[37] Jaouadi A Desouter-Lecomte M Lefebvre R and Atabek O 2013 J. Phys. B: At. Mol. Opt. Phys. 46145402
[38] Atabek O Lefebvre R Jaouadi A and Desouter-Lecomte M 2013 Phys. Rev. A 87 031403(R)
[39] Kapralova Zanska P R and Moiseyev N 2014 J. Chem. Phys. 141014307

