# Time-Dependent Schrödinger Equation with Magnetic Field

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## 1 Introduction

In this report we consider a spinless particle in  $\mathbb{R}^d$  with mass  $m \in \mathbb{R}_{\geq 0}$  and charge  $e \in \mathbb{R}$  in a homogeneous magnetic field B(t). We follow the notation introduced in the paper by Gradinaru and Rietmann and quickly recap the important elements. For a full derivation please consult [1].

#### 1.1 Mathematic model

In quantum mechanics, the time evolution of a particle subject to a magnetic field is given by the Pauli equation

$$i\hbar\partial_t \Psi(x,t) = H_P(t)\Psi(x,t) \tag{1}$$

$$H_P(t) := \frac{1}{2m} \sum_{k=1}^{d} (p_k - eA_k(x,t))^2 + e\phi(x,t) + V_{ext}(x,t)$$
 (2)

where  $V_{ext}(x,t)$  is some external potential,  $A_k(x,t)$  the k-th component of the magnetic vector potential A(x,t) and  $\phi(x,t)$  the electric potential. Because of the homogeneity of B(t) the magnetic field 2-form dA associated with B(t) is independent of x and we can rewrite the magnetic vector potential to be

$$A(x,t) := \frac{1}{2}B_{jk}(t)x^j dx^k, \tag{3}$$

where  $B(t) = (B_{jk}(t))_{j,k=1}^d$  is a real, skew-symmetric matrix. Using the operators

$$L_{ik} := x_i p_k - x_k p_i \tag{4}$$

$$H_B(t) := -\sum_{j,k=1}^{d} B_{jk}(t) L_{jk}$$
 (5)

the Pauli-Hamiltonian takes the form

$$H_P(t) = \frac{1}{2m} \left( \hbar^2(-\Delta) - e \cdot \sum_{1 \le j < k \le d} B_{jk}(t) L_{jk} + \frac{e^2}{4} \|B(t)x\|_{\mathbb{R}^d}^2 \right) + e\phi(x,t) + V_{ext}(x,t). \tag{6}$$

#### 1.2 Numerical model

We introduce the scaled Plancks constant  $\epsilon^2 := \hbar$  and redefine t, x and B to find the simplified form

$$H_P(t) = -\Delta + H_B(t) + V(x,t) \tag{7}$$

where  $V(x,t) := \frac{1}{2m} \frac{e^2}{4} ||B(t)x||_{\mathbb{R}^d}^2 + e\phi(x,t) + V_{ext}(x,t)$  can be considered to be a total effective potential. The Schrödinger equation

$$i\epsilon^2 \partial_t \Psi(x,t) = H_P(t)\Psi(x,t)$$
 (H)

is then split up into three separate parts that can be solved numerically.

$$i\epsilon^2 \partial_t \Psi = -\Delta \Psi \tag{K}$$

$$i\epsilon^2 \partial_t \Psi = H_B(t)\Psi$$
 (M)

$$i\epsilon^2 \partial_t \Psi = V(x, t)\Psi$$
 (P)

(K) can be solved discretely in Fourier-space and (P) by pointwise multiplication with  $e^{-i/\epsilon^2 \int_{t_0}^t dt V(x,t)}$ .

(M) is reduced to the linear differential equation

$$\frac{d}{dt}y(t) = B(t)y(t) \tag{B}$$

[2] proves the existence of a flow map  $U(t,t_0)$  which is a solution to (B). The unitary representation

$$\rho: SO(d) \longrightarrow U(L^2(\mathbb{R}^d)) \tag{8}$$

$$R \longmapsto (\rho(R)\Psi)(x) = \Psi(R^{-1}x) \tag{9}$$

maps the solution  $U(t,t_0)$  of (B) to a solution of (M). The proof of this statement can be found in [1]. The exact flow map  $U(t,t_0)$  can be approximated by the Magnus expansion proposed by Blanes and Moan [3]. Direct calculation shows  $[-\Delta, H_B(t)] = 0$ . Thus the flow maps solving (K) and (M) yield a solution to the differential equation

$$i\epsilon^2 \partial_t \Psi = (-\Delta + H_B(t))\Psi$$
 (K+M)

in the following way: Denote the solutions to (K) and (M) by  $\Phi_{-\Delta}$  and  $\Phi_{H_B}$  respectively. This means that the flow map to (K) satisfies

$$i\epsilon^2 \frac{\partial}{\partial t} \Phi_{-\Delta}(t, t_0) = -\Delta \Phi_{-\Delta}(t, t_0), \quad \Phi_{-\Delta}(t_0, t_0) = id, \tag{10}$$

and analogously for the flow map to (M). The flow map  $\Phi_{-\Delta+H_B}$  which is a solution to (K+M) then follows as a simple multiplication

$$\Phi_{-\Delta + H_B}(t, t_0) = \Phi_{H_B}(t, t_0)\Phi_{-\Delta}(t, t_0) = \Phi_{-\Delta}(t, t_0)\Phi_{H_B}(t, t_0). \tag{11}$$

Finally, combining the solutions to (K+M) and (P) using a splitting scheme leads to a solution of H.

## 1.3 Splitting

Consider a splitting scheme with the coefficients  $(a_i, b_i)$  for  $1 \le i \le n$  and the time grids

$$t_i = t_0 + (t - t_0) \sum_{j=1}^{i} b_j$$
 and  $s_i = t_0 + (t - t_0) \sum_{j=1}^{i} a_j$  (12)

to an initial time  $t_0$ . [1] derives an explicit expression of the solution to H as follows:

$$\Phi_{H}\left(t_{0}+Nh,t_{0}\right) \approx \left(\prod_{j=0}^{N-1}\prod_{i=0}^{n-1}\Phi_{-\Delta}\left(t_{i+1},t_{i}\right)\Phi_{\rho\left(U\left(t_{0}+Nh,t_{i}+jh\right)\right)V}\left(s_{i+1}+jh,s_{i}+jh\right)\right)\Phi_{H_{B}}\left(t_{0}+Nh,t_{0}\right)$$
(13)

where h is the splitting time step,  $\rho$  the representation defined in eq. 8 and  $U(t, t_0)$  the exact flow map to (B).

To simplify the numerical calculations the propagator for the potential part  $\Phi_V = \exp(-i\int_{t_0}^t V(x,s)ds)$  can be rewritten using the definition  $V(x,t) = ||B(t)x||_{\mathbb{R}^d}^2 + \phi(x,t) + V_{ext}(x,t)$  to

$$\int_{t_0}^t V(x,s)ds = \left\langle x, \left( -\int_{t_0}^t B^2(s)ds \right) x \right\rangle + \int_{t_0}^t \left( \phi(x,s) + V_{ext}(x,s) \right) ds. \tag{14}$$

class WaveBlocksND. FourierMagneticPropagator(parameters, potential, initial\_values)

This class can numerically propagate given initial values  $\Psi(x_0,t_0)$  on a potential hyper[source] surface V(x), in presence of a magnetic field. The propagation is done with a splitting of the time propagation operator  $\exp(-\frac{i}{\varepsilon^2}\tau H)$ . Available splitting schemes are implemented in splittingParameters.

Figure 1: Documentation-Header of the FourierMagneticPropagator Class.

## 2 Implementation

The considerations made above were implemented into the WaveBlocksND project, developed by Bourquin and Gradinaru [4]. To that avail we created a new FourierMagneticPropagator-class, based on the existing FourierPropagator-class. The full class code can be found in Appendix A. The Fourier-MagneticPropagator-class carries the header portrayed in figure 1.

In this section, we present the *postpropagate*-function which carries the implementation of section 1 and explain its approach.

The code of the *postpropagate*-function is summarised in Algorithm 1. It is based on Algorithm 1 from [1]. Its header is visible in figure 2.

## post propagate(tspan) [source]

Given an initial wavepacket  $\Psi_0$  at time t=0, calculate the propagated wavepacket  $\Psi$  at time tspan[0]. We perform  $n=\lceil tspan[0]/dt \rceil$  steps of size dt.

**Parameters:** tspan – ndarray consisting of end time at position 0, other positions are irrelevant.

Figure 2: Header of the postpropagate-function.

The flow map  $\rho$  from equation 8 effectively acts as a rotation. These rotations could not directly be applied onto the potential V or the wavefunction  $\Psi$  but had to be realised via a rotation of the underlying grid X. More precisely, the operation  $(\rho(R)A)(x) = A(R^{-1}x)$  requires us to first rotate the grid X by  $R^{-1}$  and then evaluate quantity A on the rotated grid. Because X is a class member in WaveBlocksND, after the evaluation of A on  $R^{-1} \cdot X$  the above rotation needs to be reversed to return to the original state. Otherwise subsequent evaluations would be made on the wrong grid.

Because of the rotational invariance of the first term in equation 14, in the term  $\Phi_{\rho(U(t_0+Nh,t_i+jh))V}$  only the electric potential  $\phi$  needs to be subjected to such a rotation.

#### Algorithm 1: PostPropagate Function

```
Data: As arguments to the function: class instance self; end time t.
            As arguments to the class instance: step width dt; meshgrid X; initial wave function
            \Psi_0; potential V; magnetic field B; number of components of the wavefunction N;
           scaled Planck constant \epsilon; splitting method with coefficients (a_i, b_i)_{i=1}^n; end time of
           simulation T; dimension d; frequency of writing to disk w_n.
   Result: Computes wavepacket at time t and saves it to class member. Returns: End time t.
 1 define stepwidth: n_{steps} = \lceil t/dt \rceil;
2 define time grids: t_a = t_0, t_b = t_0;
3 calculate flow map R = U(t_0 + N \cdot dt, t_0);
 4 rotate the grid by R^T;
 5 evaluate initial data on rotated grid and save to wavefunction \Psi;
 6 rotate grid back;
 7 for j = 0 to nsteps do
       for i = 0 to dim(a) do
 8
           potential propagator \Phi_V = \Phi_B \cdot \Phi_{\phi}:
 9
               calculate magnetic field propagator \Phi_B = \exp(-i\langle x, (\int_{t_a}^{t_a+a_i \cdot dt} B^2(s)ds)x\rangle);
10
               apply propagator to \Psi;
11
               rotate grid by R^T;
12
               calculate electric and external potential propagator
13
                \Phi_{\phi} = \exp(-i \int_{t_a}^{t_a + a_i \cdot dt} (\phi(x, s) + V_{ext}) ds) ;
               apply propagator to \Psi on rotated grid ;
14
               rotate grid back;
15
           calculate flow map R = R \cdot U^{-1}(t_b + b_i \cdot dt, t_b);
16
           kinetic propagator \Phi_{-\Delta}:
17
               Fast-Fourier-Transform \Psi to Fourier space;
18
               calculate the kinetic propagator Phi_{-\Delta};
19
               apply propagator to \mathcal{FFT}(\Psi);
20
               Inverse-FFT \Psi to real space;
21
           update time grids: t_a = t_a + a_i \cdot dt, t_b = t_b + b_i \cdot dt;
22
       end
23
24 end
25 return t
```

## 3 Results

In order to investigate the results of the *FourierMagneticPropagator*, we will present two examples and analyse several important metrics, mainly those of energy and norm conservation and convergence. Additionally, the time evolution is shown in appendix ??

$\mathbf{q}$	(1.0	0.0)
p	(0.0	0.0)
Q	$\begin{pmatrix} \sqrt{2.0 \cdot 0.56} \\ 0.0 \end{pmatrix}$	$ \frac{0.0}{\sqrt{2.0 \cdot 0.24}} $
Р	$ \begin{pmatrix} i/\sqrt{2.0 \cdot 0.56} \\ 0.0 \end{pmatrix} $	$\frac{0.0}{i/\sqrt{2.0\cdot0.24}}$
$\overline{S}$	(0.	.0)

Table 1: Parameters for the initial wavepacket  $\Psi_0$ .

## 3.1 Example: Threefold Morse Potential

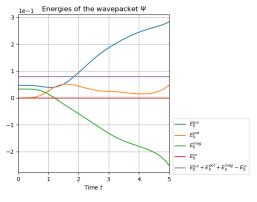
Consider the threefold morse potential for  $x \in \mathbb{R}^2$ :

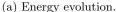
$$V_{ext}(x) = 8\left(1 - \exp\left(-\frac{\|x\|_{\mathbb{R}^2}^2}{32}(1 - \cos(3\arctan(x_2, x_1)))^2\right)\right)^2.$$
 (15)

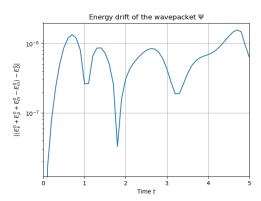
Consider the inital data  $\Psi_0 = SOMETHING$  with the parameters from table 1, step width dt = 0.01, start time  $t_0 = 0$ , end time T = 5 and the homogeneous, time-independent magnetic field  $B = \begin{pmatrix} 0 & -0.5 \\ 0.5 & 0 \end{pmatrix}$ . As the splitting method we chose Strang Splitting [6].

## 3.1.1 Energy and Norm conservation

The evolution of the energies is visible in figure 3, the evolution of the norms in figure 4. We found that both energies and norms are approximately constant.







(b) Energy drift evolution.

Figure 3: Energy and energy drift.

#### 3.1.2 Resource Consumption

We ran the simulation on a CPU consisting of 2x AMD Opteron(tm) Processor 6174 with 24 Cores and recorded the resource consumption using Linux' pidstat and time commands. The simulation lasted a total of 5:50:44 (h:min:s) consuming on average 1068% CPU. The usage of RSS, VSZ and CPU over time is visible in figure 5.

#### 3.1.3 Convergence

## 3.2 Example: Torsional Potential

Consider the torsional potential from [5] for  $x \in \mathbb{R}^2$ :

$$V_{ext}(x) = \sum_{i=1}^{2} 1 - \cos(x_i). \tag{16}$$

//applications of this potential?

Consider  $\Psi_0$  to be as in the example above with the parameters from table 1 and the splitting to be Strang Splitting.

#### 3.2.1 Energy and Norm conservation

The energy and norm evolution over time is depicted in figures 6 and ?? respectively.

#### 3.2.2 Time Evolution

#### 3.2.3 Resource Consumption

We ran the simulation on a setup identical to the one mentioned above and again recorded the resource consumption using Linux' time and pidstat commands. The simulation lasted for a total of 7:45:18 (h:min:s) and consumed 247% CPU. The usage of RSS, VSZ and CPU over time is visible in figure 7.

#### 3.2.4 Convergence

## 4 Summary and Conclusion

# Appendices

## A Code

```
r"""The WaveBlocks Project

This file contains the Fourier Magnetic Propagator class. The wavefunction
math: '\Psi' is propagated in time with a splitting of the
exponential :math: '\exp(-\frac{i}{\varepsilon^2} \tau H)'.

Quuthor: R. Bourquin
```

```
8 @copyright: Copyright (C) 2012, 2016 R. Bourquin
9 @license: Modified BSD License
10 """
12 from numpy import array, complexfloating, dot, exp, eye, zeros, shape
13 from numpy.fft import fftn, ifftn
14 from scipy.linalg import expm
15
16 from WaveBlocksND.BlockFactory import BlockFactory
17 from WaveBlocksND.Propagator import Propagator
18 from WaveBlocksND.KineticOperator import KineticOperator
19 from WaveBlocksND.MagneticField import MagneticField
20 from WaveBlocksND.SplittingParameters import SplittingParameters
22 __all__ = ["FourierMagneticPropagator"]
23
24
  class FourierMagneticPropagator(Propagator, SplittingParameters):
25
      r"""This class can numerically propagate given initial values :math: `\Psi(x_0, t_0)
26
      ) ' on
      a potential hyper surface :math: `V(x)`, in presence of a magnetic field. The
27
      propagation is done with a splitting
      of the time propagation operator :math: '\exp(-\frac{i}{\pi c^2} \times H)'.
      Available splitting schemes are implemented in :py:class:'SplittingParameters'.
29
30
31
      def __init__(self, parameters, potential, initial_values):
32
           r"""Initialize a new :py:class:'FourierMagneticPropagator' instance.
33
      Precalculate the
          the kinetic operator :math:'T_e' and the potential operator :math:'V_e'
          used for time propagation.
35
36
          :param parameters: The set of simulation parameters. It must contain at least
37
                              the semi-classical parameter :math: '\varepsilon' and the
38
                              time step size :math: '\tau'.
39
          :param potential: The potential :math:'V(x)' governing the time evolution.
40
41
           :type potential: A :py:class:'MatrixPotential' instance.
          :param initial_values: The initial values :math: '\Psi(\Gamma, t_0)' given
42
                                  in the canonical basis.
43
          :type initial_values: A :py:class:'WaveFunction' instance.
44
45
           :raise: :py:class:'ValueError' If the number of components of :math:'\Psi'
46
      does not match the
                              number of energy surfaces :math:'\lambda_i(x)' of the
47
      potential.
48
           :raise: :py:class:'ValueError' If the number of components of :math:'\Psi'
49
      does not match the dimension of the magnetic field :math: ' \vee C(B)(x)'.
50
           :raise: :py:class:'ValueError' If the dimensions of the splitting scheme
51
      parameters :math: 'a' and :math: 'b' are not equal.
          # The embedded 'MatrixPotential' instance representing the potential 'V'.
53
          self._potential = potential
54
55
           # The initial values of the components '\psi_i' sampled at the given grid.
56
57
          self._psi = initial_values
58
```

```
if self._potential.get_number_components() != self._psi.get_number_components
       ():
               raise ValueError("Potential dimension and number of components do not
60
       match.")
61
           # The time step size.
62
           self._dt = parameters["dt"]
63
64
65
           # Final time.
           self._T = parameters["T"]
66
67
           # The model parameter '\varepsilon'.
68
69
           self._eps = parameters["eps"]
70
           # Spacial dimension d
71
72
           self._dimension = parameters["dimension"]
73
           # The position space grid nodes '\Gamma'.
74
           self._grid = initial_values.get_grid()
75
76
           # The kinetic operator 'T' defined in momentum space.
77
           self._KO = KineticOperator(self._grid, self._eps)
78
79
           # Exponential '(\exp(-i/2*eps^2*dt*T)') used in the Strang splitting.
80
81
           # not used
           self._KO.calculate_exponential(-0.5j * self._dt * self._eps**2)
82
           self._TE = self._KO.evaluate_exponential_at()
83
84
           # Exponential '\exp(-i/eps^2*dt*V)' used in the Strang splitting.
85
86
           # not used
           self._potential.calculate_exponential(-0.5j * self._dt / self._eps**2)
87
           VE = self._potential.evaluate_exponential_at(self._grid)
88
89
           self._VE = tuple([ve.reshape(self._grid.get_number_nodes()) for ve in VE])
90
           # The magnetic field
91
           self._B = MagneticField(parameters["B"])
92
93
           # check if magnetic field and potential are of same dimension
           if self._B.get_dimension() != self._dimension:
94
               raise ValueError("Spacial dimension of potential and magnetic field must
95
       be the same")
96
           #precalculate the splitting needed
97
           self._a, self._b = self.build(parameters["splitting_method"])
98
           if shape(self._a) != shape(self._b):
99
               raise ValueError("Splitting scheme shapes must be the same")
101
           # Get inital data as function
           packet_descr = parameters["initvals"][0]
103
           self._initalpacket = BlockFactory().create_wavepacket(packet_descr)
104
105
106
       # TODO: Consider removing this, duplicate
       def get_number_components(self):
108
           r"""Get the number :math:'N' of components of :math:'\Psi'.
109
           :return: The number :math:'N'.
111
112
           return self._potential.get_number_components()
```

```
114
       def get_wavefunction(self):
116
           r"""Get the wavefunction that stores the current data :math:'\Psi(\Gamma)'.
117
118
           :return: The :py:class:'WaveFunction' instance.
119
120
           return self._psi
121
122
123
124
       def get_operators(self):
           r"""Get the kinetic and potential operators :math:'T(\omega)' and :math:'V(\omega)
125
       Gamma) '.
126
127
           :return: A tuple :math: '(T, V)' containing two ''ndarrays''.
128
           # TODO: What kind of object exactly do we want to return?
129
           self._KO.calculate_operator()
130
           T = self._KO.evaluate_at()
131
           V = self._potential.evaluate_at(self._grid)
132
133
           V = tuple([v.reshape(self._grid.get_number_nodes()) for v in V])
           return (T, V)
134
135
136
       @staticmethod
137
       def _Magnus_CF4(tspan, B, N, *args):
138
           r"""Returns the Fourth Order Magnus integrator :math: '\Omega(A)' according to
139
        [#]_.
140
141
           :param tspan: Full timespan of expansion.
142
           :param B: Magnetic field matrix :math: (B(t) = (B_{j,k}(t))_{1 \leq j,k} (t)
143
       }'.
144
           :param N: Number of timesteps for the expansion.
145
146
147
           :param *args: Additional arguments for the magnetic field :math: 'B(t, *args)'
148
           .. [#] S. Blanes and P.C. Moan. "Fourth- and sixth-order commutator-free
149
       Magnus integrators for linear and non-linear dynamical systems". Applied Numerical
       Mathematics, 56(12):1519 - 1537, 2006.
           0.00
150
           # Magnus constants
           c1 = 0.5*(1.0 - 0.5773502691896258)
           c2 = 0.5*(1.0 + 0.5773502691896258)
           a1 = 0.5*(0.5 - 0.5773502691896258)
154
           a2 = 0.5*(0.5 + 0.5773502691896258)
156
           R = 1.*eye( len( B(1.*tspan[0], *args) ) )
157
           h = (tspan[1]-tspan[0]) / (1.*N)
158
           for k in range(N):
159
                t0 = k*h + tspan[0]
               t1 = t0 + c1*h
161
               t2 = t0 + c2*h
162
163
               B1 = B(t1, *args)
               B2 = B(t2, *args)
164
               R = dot(expm(a1*h*B1+a2*h*B2), dot(expm(a2*h*B1+a1*h*B2), R))
166
```

```
return R
168
169
       def post_propagate(self, tspan):
170
           r"""Given an initial wavepacket :math: '\Psi_0' at time :math: 't=0', calculate
       the propagated wavepacket :math: '\Psi' at time :math: 'tspan \[ 0 \]'. We perform :
       172
           :param tspan: :py class:'ndarray' consisting of end time at position 0, other
173
       positions are irrelevant.
174
175
176
           # (ignoriere tspan[0])
           nsteps = int(tspan[0] / self._dt + 0.5)
177
178
           print("Perform " + str(nsteps) + " steps from t = 0.0 to t = " + str(tspan[0])
179
180
           # Magnetfeld Matrix B(t)
181
           B = lambda t: self._B(t)
182
183
           #how many components does Psi have
184
185
           N = self._psi.get_number_components()
186
           \#start time t_0 = 0?
187
           t0 = 0
188
           t_a = t0
189
           t_b = t0
190
191
192
           #calculate R = U(t0 + N*h, t0)
           \# Use \ \mathbb{N} = n\_steps to account for large time difference
193
           t_interval = array([t0, tspan[0]])
194
195
           R = FourierMagneticPropagator._Magnus_CF4(t_interval, B, nsteps)
196
           \mbox{\tt\#} rotate the grid by the transpose of R
197
           self._grid.rotate(R.T)
198
199
           # Compute rotated initial data
200
           X = self._grid.get_nodes(flat=True)
201
202
           values = self._initalpacket.evaluate_at(X, prefactor=True)
           values = tuple([val.reshape(self._grid.get_number_nodes()) for val in values])
203
           self._psi.set_values(values)
204
205
           self._grid.rotate(R)
206
207
           #calculate the necessary timesteps
208
           for j in range(nsteps):
209
               for i in range(len(self._a)):
210
                   # Integral -\int_{tspan}[0]^{tspan}[1]B^2(s)ds und zugehA^{rige}
211
       Propagation
                   # (siehe Paper, Remark 3.1)
212
                   minus_B_squared = lambda t: (-1.0) * dot(B(t), B(t))
213
                   A = 1.0 / 8.0 * MagneticField.matrix_quad(minus_B_squared, t_a, t_a +
214
       self._a[i]*self._dt)
215
                   X = self._grid.get_nodes(flat=True)
216
217
                   VB = sum(X * dot(A, X))
                   VB = VB.reshape(self._grid.get_number_nodes())
218
```

```
prop = \exp(-1.0j / \text{self.\_eps**2 * VB}) # ev. -0.5j durch -1j ersetzen
220
                    values = self._psi.get_values()
221
                    values = [prop * component for component in values]
222
223
                    self._potential.calculate_exponential(-1.0j * self._a[i]*self._dt /
224
       self._eps**2)
225
                    self._grid.rotate(R.T)
226
                    VE = self._potential.evaluate_exponential_at(self._grid)
227
                    self._VE = tuple([ve.reshape(self._grid.get_number_nodes()) for ve in
228
       VE])
229
230
                    #apply it
231
                    values = [self._VE * component for component in values]
                    self._grid.rotate(R)
232
233
                    t_interval[0] = t_b
234
                    t_interval[1] = t_b + self._b[i]*self._dt
235
236
                    U = (FourierMagneticPropagator._Magnus_CF4(t_interval, B, 1)).T
237
238
                    R = dot(R, U)
                    if(R.shape != U.shape):
239
                        raise ValueError("Shapes of R and U do not match")
240
241
                    #check for obsolete splitting steps
242
                    if(self._b[i] != 0):
243
                        values = [fftn(component) for component in values]
244
245
                        # Apply the kinetic operator
246
                        self._KO = KineticOperator(self._grid, self._eps)
247
                        self._KO.calculate_exponential(-0.5j * self._eps**2 * self._b[i]*
248
       self._dt)
249
                        TE = self._KO.evaluate_exponential_at()
250
251
                        values = [TE * component for component in values]
252
                        # Go back to real space
253
254
                        values = [ifftn(component) for component in values]
255
                    #Apply
256
                    self._psi.set_values(values)
257
258
259
                    #update t_a and t_b
                    t_a = t_a + self._a[i]*self._dt
260
                    t_b = t_b + self._b[i]*self._dt
261
262
           return tspan[0]
263
264
265
266
       def propagate(self, tspan):
           r"""This method does nothing.
267
268
```

## B Time evolution

#### **B.1** Threefold Morse Potential

The time evolution of the wavepacket  $\Psi$  is portrayed in figure 8. At a position x the color encodes the phase of  $\Psi(x)$ , the brightness of the pixel encodes the intensity  $|\Psi(x)|$ .

#### **B.2** Torsional Potential

The time evolution of the wavepacket  $\Psi$  is portrayed in figure 9. At a position x the color encodes the phase of  $\Psi(x)$ , the brightness of the pixel encodes the intensity  $|\Psi(x)|$ .

## References

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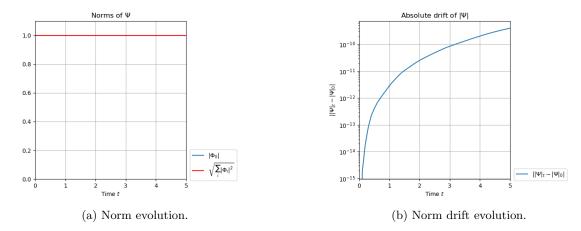
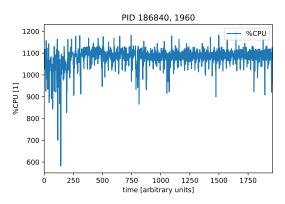
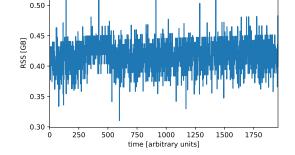


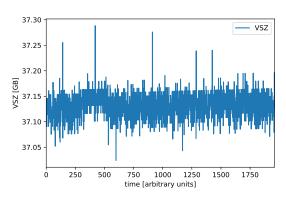
Figure 4: Norm and norm drift.





(a) Percentage of CPU used by the simulation.

(b) Amount of RSS (Resident Set Size) used by the simulation.



(c) Amount of VSZ (Virtual Memory Size) used by the simulation.

Figure 5: Resource consumption of the simulation with the threefold morse potential.

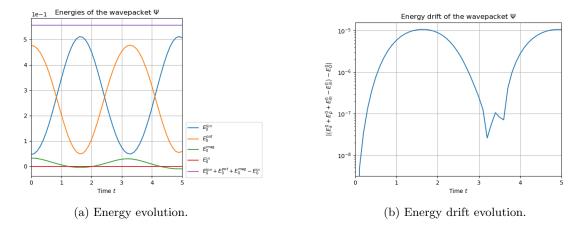
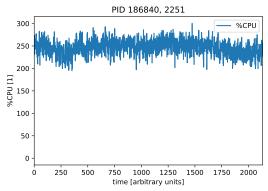
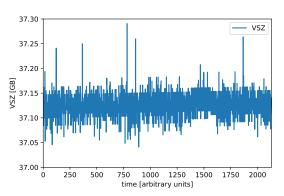


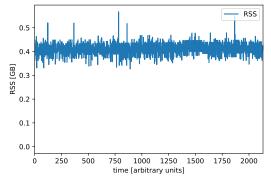
Figure 6: Energy and energy drift.







(c) Amount of VSZ (Virutal Memory Size) used by the simulation.



(b) Amount of RSS (Resident Set Size) used by the simulation.

Figure 7: Resource consumption of the simulation with the torsional potential. The difference in CPU percentage between the torsional potential and the morse potential is mostly due to different loads on the machines at the time of simulation.

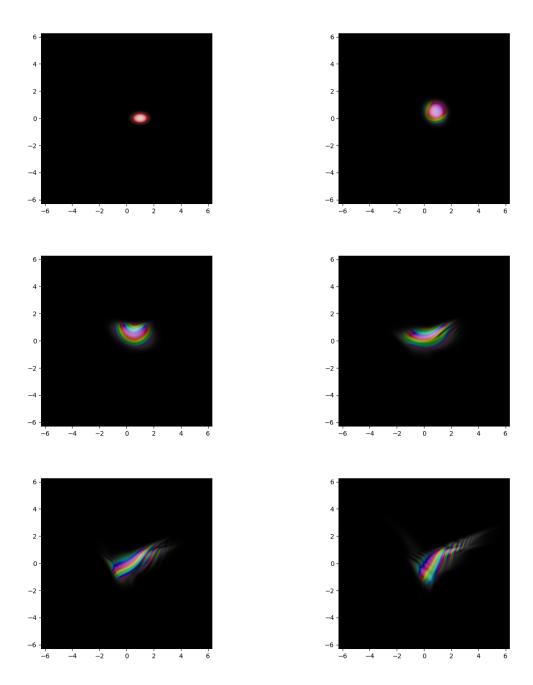


Figure 8: Time evolution of the wave packet  $\Psi$  with initial data  $\Psi_0$  in the threefold morse potential and a homogeneous, time-independent magnetic field.

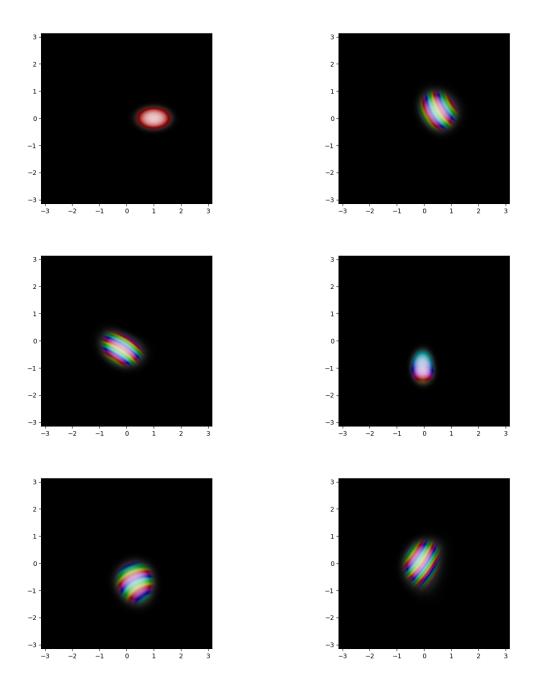


Figure 9: Time evolution of the wave packet  $\Psi$  with initial data  $\Psi_0$  in the torsional potential and a homogeneous, time-independent magnetic field.