Solution of the Time-dependent Schrödinger Equation using the Crank-Nicolson algorithm with MPI on a 2-D regular Cartesian grid

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Time-dependent Schrödinger Equation

Schrödinger Equation in 2-D for an electron in a given potential

$$i\hbar \frac{\partial}{\partial t} |\psi(t, \mathbf{x})\rangle = \hat{H}(\hat{\mathbf{p}}, \hat{\mathbf{x}}) |\psi(t, \mathbf{x})\rangle, \quad \hat{H} = \frac{\hat{p}_x^2}{2m} + \frac{\hat{p}_y^2}{2m} + \hat{V}(x, y)$$
 (1)

Attempt to simply discretize the differential operators,

$$\partial_t \psi(t, \mathbf{x}) pprox \frac{\psi_{j,l}^{n+1} - \psi_{j,l}^n}{\Delta t},$$

$$\frac{\hat{\boldsymbol{p}}^2}{2m}\psi(t,\boldsymbol{x}) = -\frac{\hbar^2}{2m}\nabla^2\psi(t,\boldsymbol{x}) \approx -\frac{\hbar^2}{2m}\frac{\psi_{j+1,l}^{n+1} - 2\psi_{j,l}^{n+1} + \psi_{j-1}^{n+1}}{(\Delta x)^2} + \dots$$

is – as the von Neumann stability analysis shows – unconditionally stable. However, each solution of Schrödinger's equation has to fulfill a continuity equation for the probability $w(t, \mathbf{x}) \, \mathrm{d}^3 x = \psi^* \psi \, \mathrm{d}^3 x$ to find the electron within a volume $\mathrm{d}^3 x$ around \mathbf{x} ,

$$\partial_t w(t, \mathbf{x}) + \nabla \cdot \mathbf{j} = 0, \quad \mathbf{j} = \frac{\hbar}{2\mathrm{i}m} [\psi^*(\nabla \psi) - (\nabla \psi^*)\psi]$$
 (2)

From the continuity equation follows the conservation of the norm of the wave function:

$$\frac{\mathrm{d}}{\mathrm{d}t}\int\,\mathrm{d}^3x\,w(t,\textbf{\textit{x}})=\int\,\mathrm{d}^3x\,\partial_tw(t,\textbf{\textit{x}})=-\int\,\mathrm{d}^3x\,\nabla\cdot\textbf{\textit{j}}(t,\textbf{\textit{x}})=-\int\,\mathrm{d}\textbf{\textit{S}}\cdot\textbf{\textit{j}}(t,\textbf{\textit{x}})$$

$$\int_{\mathbb{R}^{3}} d^{3}x \|w(0,x)\|^{2} = 1 \xrightarrow{SE} \int_{\mathbb{R}^{3}} d^{3}x \|w(t,x)\|^{2} = 1$$
 (3)

The normalization of the wave function $\psi(t,x)$ is however not conserved by the fully implicit discretization as this method is not unitary \Rightarrow a discretization preserving the hermiticity of the Hamiltonian is needed that maintains the normalization of the wavefunction.

The temporal evolution of the wave function can be expressed using the unitary time-evolution operator $\hat{U}(t)$:

$$|\psi(t+\Delta t, \mathbf{x})\rangle \simeq \hat{U}(t) |\psi(t, \mathbf{x})\rangle, \qquad \hat{U}(t) = \exp(-\mathrm{i}\hat{H}t/\hbar)$$
 (4)

With the time-evolution operator one can find the explicit Forward-Time Centered-Space (FTCS) scheme $\psi_{j,l}^{n+1}=(1-\mathrm{i}\hat{H}\Delta t)\psi_{j,l}^n$ and the implicit Backward-Time scheme $(1+\mathrm{i}\hat{H}\Delta t)\psi_{j,l}^{n+1}=\psi_{j,l}^n$. Neither scheme is unitary and thus violates probability conservation.

A finite-difference approximation of $\exp(-i\hat{H}\Delta t/\hbar)$ that is unitary was found by Cayley:

$$\exp(-\mathrm{i}\hat{H}\Delta t/\hbar) = rac{1-rac{\mathrm{i}H\Delta t}{2\hbar}}{1+rac{\mathrm{i}\hat{H}\Delta t}{2\hbar}} + \mathcal{O}((\Delta t)^2)$$

$$(4) \Rightarrow \left(1 + \frac{\mathrm{i}}{2\hbar} \hat{H} \Delta t\right) |\psi(t + \Delta t, \mathbf{x})\rangle = \left(1 - \frac{\mathrm{i}}{2\hbar} \hat{H} \Delta t\right) |\psi(t, \mathbf{x})\rangle$$
 (5)

This is exactly the Crank-Nicolson scheme for the Schrödinger equation.

Discretization

By substituting the 2-D discrete wave function $\psi(t; x, y) \to \psi_{i,l}^n$ into Eq. (5) and considering the Hamiltonian in the form

$$\hat{H} = |\hat{\boldsymbol{\rho}}| = -i\hbar\nabla| = -\frac{\hbar^2}{2m}\frac{\partial^2}{\partial x^2} - \frac{\hbar^2}{2m}\frac{\partial^2}{\partial y^2} + \hat{V}(x,y)$$

one obtains the discretized 2-D Schrödinger Equation

$$\psi_{j,l}^{n+1} - \frac{i\Delta t}{2\hbar} \left(\frac{\hbar^2}{2m} \frac{\psi_{j+1,l}n + 1 - 2\psi_{j,l}^{n+1} + \psi_{j-1,l}^{n+1}}{(\Delta x)^2} + \frac{\psi_{j,l+1}^{n+1} - 2\psi_{j,l}^{n+1} + \psi_{j,l-1}^{n+1}}{(\Delta y)^2} - V_{j,l}\psi_{j,l}^{n+1} \right)$$

$$= \psi_{j,l}^n + \frac{i\Delta t}{2\hbar} \left(\frac{\hbar^2}{2m} \frac{\psi_{j+1,l}^n - 2\psi_{j,l}^n + \psi_{j-1,l}^n}{(\Delta x)^2} + \frac{\hbar^2}{2m} \frac{\psi_{j,l+1}^n - 2\psi_{j,l}^n + \psi_{j,l-1}^n}{(\Delta y)^2} - V_{j,l}\psi_{j,l} \right)$$
(6)

Rewrite Eq. (6) in matrix-form. Assign a short-hand notation for the prefactors of the wave function at the next time step at various spatial grid points:

$$\begin{split} \psi_{j,l}^{n+1} : & b_{j,l} = 1 + \frac{\mathrm{i}\Delta t}{\hbar} \frac{\hbar^2}{2m} \left(\frac{1}{(\Delta x)^2} + \frac{1}{(\Delta y)^2} \right) + \frac{\mathrm{i}\Delta t}{2\hbar} V_{j,l} \\ \psi_{j-1,l}^{n+1} : & c = -\frac{\mathrm{i}\Delta t}{2\hbar} \frac{\hbar^2}{2m} \frac{1}{(\Delta x)^2} \\ \psi_{j+1,l}^{n+1} : & a = c \\ \psi_{j,l+1}^{n+1} : & d = -\frac{\mathrm{i}\Delta t}{2\hbar} \frac{\hbar^2}{2m} \frac{1}{(\Delta x)^2} \\ \psi_{j,l-1}^{n+1} : & e = d \end{split}$$

and for the RHS:

$$\begin{split} \psi_{j,l}^n: & \qquad f_{j,l} = 1 - \frac{\mathrm{i}\Delta t}{\hbar} \frac{\hbar^2}{2m} \left(\frac{1}{(\Delta x)^2} + \frac{1}{(\Delta y)^2} \right) - \frac{\mathrm{i}\Delta t}{2\hbar} V_{j,l} \\ \psi_{j+1,l}^n: & \qquad g = \frac{\mathrm{i}\Delta t}{2\hbar} \frac{\hbar^2}{2m} \frac{1}{(\Delta x)^2} \\ \psi_{j-1,l}^n: & \qquad h = g \\ \psi_{j,l+1}^n: & \qquad k = \frac{\mathrm{i}\Delta t}{2\hbar} \frac{\hbar^2}{2m} \frac{1}{(\Delta y)^2} \\ \psi_{l,l-1}^n: & \qquad p = k \end{split}$$

LHS of the discretized Schrödinger Equation

Map the two indices j and l of the wave function $\psi_{j,l}^n$ on a single one $(j,l) \mapsto i$:

$$i = 1 + (j-1)(L-1) + l - 1$$
 for $j = 0, 1, ..., J$; $l = 0, 1, ..., L$

Thereby $\psi_{1,1}^n \to \psi_1^n, \ \psi_{1,2}^n \to \psi_2^n, \dots, \ \psi_{1,J}^n \to \psi_J^n, \ \psi_{2,1}^n \to \psi_{J+1}^n, \dots$ For simplicity (and lack of space) shown here only for J=L=4, though later $J,L\sim 10^3$ will be used. It exhibits sub-blocks of shape $J\times L$:

$$\begin{pmatrix} b_{1,1} & d & 0 & a & 0 & 0 & 0 & 0 & 0 \\ e & b_{1,2} & d & 0 & a & 0 & 0 & 0 & 0 & 0 \\ 0 & e & b_{1,3} & 0 & 0 & a & 0 & 0 & 0 & 0 \\ c & 0 & 0 & b_{2,1} & d & 0 & a & 0 & 0 & 0 \\ 0 & c & 0 & e & b_{2,2} & d & 0 & a & 0 & 0 \\ 0 & 0 & c & 0 & e & b_{2,2} & d & 0 & a & 0 \\ 0 & 0 & c & 0 & e & b_{2,3} & 0 & 0 & a \\ 0 & 0 & 0 & c & 0 & e & b_{3,1} & d & 0 \\ 0 & 0 & 0 & 0 & c & 0 & e & b_{3,2} & d \\ 0 & 0 & 0 & 0 & 0 & c & 0 & e & b_{3,3} \end{pmatrix} \begin{pmatrix} \psi_1^{n+1} \\ \psi_2^{n+1} \\ \psi_3^{n+1} \\ \psi_5^{n+1} \\ \psi_7^{n+1} \\ \psi_8^{n+1} \\ \psi_8^{n+1} \\ \psi_8^{n+1} \\ \psi_9^{n+1} \end{pmatrix}$$

RHS of the discretized Schrödinger Equation

Vectorization of the RHS is not necessary when working with Fortran exclusively but essential when using the numpy module with Python:

$$\mathbf{r} = \begin{pmatrix} f_{1,1} & k & 0 & g & 0 & 0 & 0 & 0 & 0 \\ p & f_{1,2} & k & 0 & g & 0 & 0 & 0 & 0 \\ 0 & p & f_{1,3} & 0 & 0 & g & 0 & 0 & 0 \\ h & 0 & 0 & f_{2,1} & k & 0 & g & 0 & 0 \\ 0 & h & 0 & p & f_{2,2} & k & 0 & g & 0 \\ 0 & 0 & h & 0 & p & f_{2,3} & 0 & 0 & g \\ 0 & 0 & 0 & h & 0 & 0 & f_{3,1} & k & 0 \\ 0 & 0 & 0 & 0 & h & 0 & p & f_{3,2} & k \\ 0 & 0 & 0 & 0 & 0 & h & 0 & p & f_{3,3} \end{pmatrix} \begin{pmatrix} \psi_1^n \\ \psi_2^n \\ \psi_3^n \\ \psi_4^n \\ \psi_5^n \\ \psi_7^n \\ \psi_8^n \end{pmatrix}$$

After vectorization, the discretized Schrödinger Equation simplifies to

$$\hat{A}\psi^{n+1} = \mathbf{r} = \hat{B}\psi^n$$

Pseudo-code:

for n in range(N):

- ullet Compute $\hat{B}\psi^n=oldsymbol{r}$ as a matrix-vector product
- Solve $\hat{A}\psi^{n+1} = r$ with LU decomposition
- ullet This yields ψ^{n+1}
- Map the rolling index i back to indices j, l: $\psi_i^{n+1} \to \psi_{j,l}^{n+1}$
- Check that the wave packet remains normalized:

$$\int_{\Delta^2} d^2x \|\psi(t, \boldsymbol{x})\|^2 \approx \sum_{j,l=1}^{J,L} \Delta x \Delta y \|\psi_{j,l}^n\|^2 \stackrel{!}{=} 1$$

Setting up the matrices in Python

```
# Build the main diagonal
b_main_diag = [] # coefficient matrix of LHS
f_main_diag = [] # coefficient matrix of RHS
for j in range(J-1):
   for 1 in range(J-1):
        b_main_diag.append(b[j,1]); f_main_diag.append(f[j,1])
e_minor_diag = np.ones((len(b_main_diag)-1,), dtype=complex)*e
d_minor_diag = np.ones((len(b_main_diag)-1,), dtype=complex)*d
a_minor_diag = np.ones((len(b_main_diag)-3,), dtype=complex)*a
c_minor_diag = np.ones((len(b_main_diag)-3,), dtype=complex)*c
k_minor_diag = np.ones((len(f_main_diag)-1,), dtype=complex)*k
p_minor_diag = np.ones((len(f_main_diag)-1,), dtype=complex)*p
g_minor_diag = np.ones((len(f_main_diag)-3,), dtype=complex)*g
h_minor_diag = np.ones((len(f_main_diag)-3,), dtype=complex)*h
# Build the coefficient matrices for A*psi^(n+1) = B*psi^n
A = np.diag(b_main_diag) + np.diag(e_minor_diag,-1) + np.diag(d_minor_diag,1) \
   + np.diag(c_minor_diag,-3) + np.diag(a_minor_diag,3)
B = np.diag(f_main_diag) + np.diag(p_minor_diag,-1) + np.diag(k_minor_diag,1) \
   + np.diag(h_minor_diag,-3) + np.diag(g_minor_diag,3)
                                                    4 D > 4 A > 4 B > 4 B > B 9 Q Q
```

Setting-up the RHS in Python

```
for n in range(Nt):
    if n == 0:
        psi_n_mat = psi_0
        # Rewrite matrix psi_jl as a vector psi_i
        psi_n_vec = np.empty(((J-1)**2,), dtype=complex)
        for j in range(J-1):
            for l in range(J-1):
                i = ind(i,1)
                psi_n_vec[i] = psi_n_mat[j,1]
    else:
        psi_n_vec = psi_np1_vec
    RHS = np.dot(B,psi_n_vec)
    r = R.H.S
```

LU decomposition; finally to be parallelized as Fortran code

```
# LU decomposition:
P, L, U = scipy.linalg.lu(A)
LU = L + U
nA = np.size(A, 0)
y = np.zeros((nA,), dtype=complex)
x = np.zeros((nA,), dtype=complex)
r = np.dot(np.linalg.inv(P),r)
for i in range(1, nA+1):
   s = r[i-1]
   for j in range(1, i):
       s = s - LU[i-1, j-1]*v[j-1]
   v[i-1] = s
for i in range(nA, 0, -1):
    s = v[i-1]
    for j in range(i+1, nA+1):
        s = s - LU[i-1, j-1] *x[j-1]
    x[i-1] = s/LU[i-1,i-1]
```

 $psi_np1_vec = x$

Parallelizing the code

The following strategies for code parallelization systems are envisaged:

- Application of Distributed SuperLU (SuperLU_DIST) of the SuperLU library
- Crank-Nicolson predictor-corrector (CNPC) method
 - Decomposition of spatial domain Ω in subdomains Ω_i which can be assigned to their own processor
 - Forward Euler to predict unknown values on the interface Γ between subdomains $\Omega_i \subset \Omega$
 - Solving the problem independently on the subdomains Ω_i with Backward Euler using the predicted values ψ_T^* at t_{n+1}
 - \bullet Correction of interface values ψ_{\varGamma}^{n+1} with Backward Euler using interior values