

Project 2

Quantum Mechanics

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Finite Differences Method

For all time-dependent problems e.g., superposition states or time-dependent potentials, I use the finite difference in the time domain (FDTD) method to solve Schrodinger's equation. The algorithm is well-known, so I will just sketch the major result I use here. The idea is to break up the time-dependent Schrodinger equation into coupled differential equations for the real and imaginary parts of the wavefunction.

Let ℓ be a discrete spatial coordinate and m, n index time for the real $\psi_R(\ell)$ and imaginary $\psi_I(\ell)$ parts of the wavefunction, respectively. The update equations are

$$\frac{1}{\Delta\tau} (\psi_I^{n+1}(\ell) - \psi_I^n(\ell)) = \eta (\psi_R^m(\ell + 1) - 2\psi_R^m(\ell) + \psi_R^m(\ell - 1)) - V(\ell)\psi_R^m(\ell)$$

$$\frac{1}{\Delta\tau} (\psi_R^{m+1}(\ell) - \psi_R^m(\ell)) = -\eta (\psi_I^n(\ell + 1) - 2\psi_I^n(\ell) + \psi_I^n(\ell - 1)) - V(\ell)\psi_I^n(\ell)$$

where $\tau = t/\hbar$ and $\eta = \frac{\hbar^2}{2m\Delta x^2}$ (and we set $\eta = 1$) is the hopping parameter. You solve the system numerically by alternatively updating the real and imaginary wavefunctions in time. Suppose the initial state is $|\alpha\rangle = \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle)$ where $|0\rangle$ and $|1\rangle$ are the ground state and first excited state for the infinite square well. Then we find the position representation $\langle i|0\rangle$ and $\langle i|1\rangle$ by solving the time-independent Schrodinger equation as in Project 1. We then can construct $\langle i|\alpha\rangle$, which is purely real, and assign this as the initial condition $\psi_R^0(\ell)$.

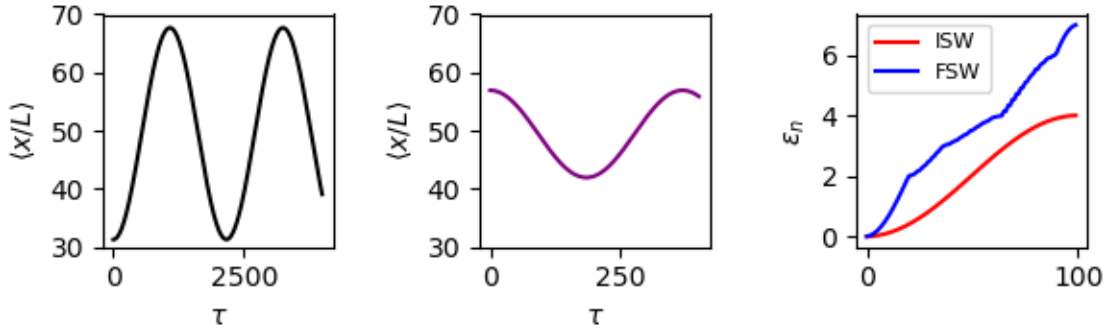


Figure 1: Expectation values of position as a function of time for the infinite (left) and finite (middle) square well, plus the eigenvalue spectrum for both (right). $d\tau = 0.01, \eta = 1, N_x = 100, N_t = 4 \times 10^5$

Part 1

Answers

(A) Since the wavefunction is a superposition of eigenkets, I expect the probability density to change in time, probably in an oscillatory fashion.

(B) The time scale should depend on the energies of the eigenkets of the superposition

(C) See Figure 1 for FDTD simulation of the infinite and finite square wells

(D) For any superposition state, we can write

$$\langle x \rangle = \sum_{a'} \sum_{a''} c_{a'}^* c_{a''} \langle a' | x | a'' \rangle \exp \left(\frac{-i(E_{a''} - E_{a'})t}{\hbar} \right)$$

which in this particular case (using $\tau = t/\hbar$) is

$$\langle x \rangle = \frac{1}{2} (\langle 0 | x | 1 \rangle \exp(i\Delta E \tau) + \langle 1 | x | 0 \rangle \exp(-i\Delta E \tau))$$

The angular frequency will be higher for the finite square well because $\Delta E = E_1 - E_0$ is larger see the derivative of the eigenvalue spectrum in Figure 1).

(E) See email attachments for the gif animation corresponding to the simulations in Figure 1

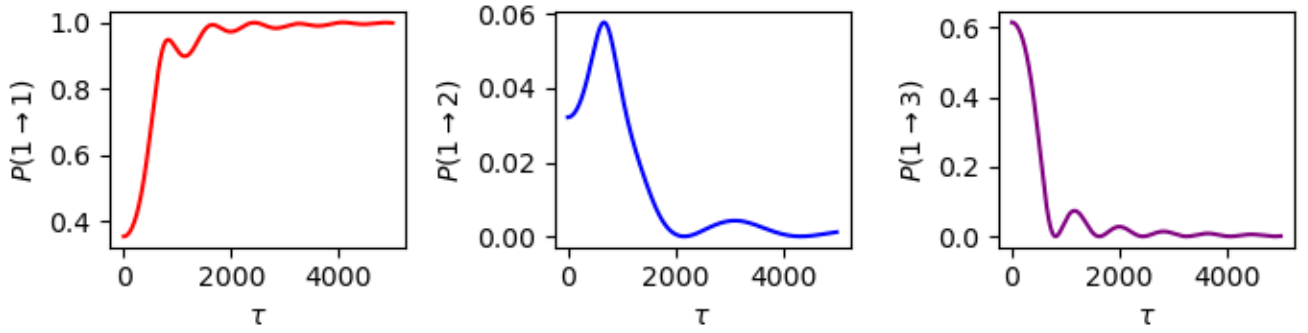


Figure 2: Transition probability $P(1 \rightarrow 1) = |c_1(t)|^2$ (left) computed with Equation 2, $P(1 \rightarrow 2) = |c_2(t)|^2$ (middle), and $P(1 \rightarrow 3) = |c_3(t)|^2$ (right), both computed with Equation 1. Expansion coefficients were renormalized at each time step. $d\tau = 0.1$, $N_t = 5 \times 10^4$

Part 2

Time-dependent perturbation theory for a constant perturbation

We are using the time-dependent Hamiltonian

$$H(x, t) = H_0(x) + \lambda(1 - e^{-t/\tau})V(x)$$

where $V(x)$ is the finite square well potential. We are assuming that $\tau \rightarrow \infty$ so the potential turns on exactly at $t = 0$, giving a **constant perturbation**.

Answers

(F) If $\lambda = 0$, we do not expect $\langle x/L \rangle$ to oscillate because $H(t) = H_0$ and $|\psi\rangle = |i\rangle$ which is an eigenstate of H_0 .

(G) If $\lambda \neq 0$, we do expect $\langle x/L \rangle$ to oscillate because turning on the potential $V(x)$ makes $|\psi\rangle$ a superposition of eigenkets of the $H(t) = H_0 + \lambda V(x)$

(H)

Now, we are told that we start in the ground state of H_0 , which is the Hamiltonian for an infinite square well, i.e. $|\psi(0)\rangle = |i\rangle = |1\rangle$. recall that in time-dependent first order perturbation theory, Schrodinger's equation becomes

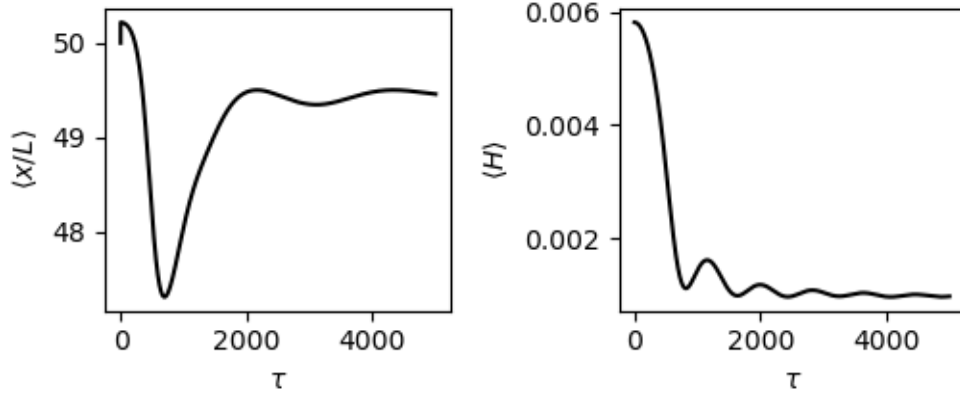


Figure 3: Analytical calculation of the expectation value of position (left) and energy (right)

$$i\hbar \frac{d}{dt} c_n^{(0)}(t) = 0$$

$$i\hbar \frac{d}{dt} c_n^{(1)}(t) = \sum_i V_{ni} e^{i\omega_{ni}t} c_i^{(0)}(t)$$

where $V_{ni} = \langle n|V|i\rangle$ and $c_n(t)$ is the expansion coefficient for **unperturbed** energy eigenket $|n\rangle$. In general, expansion coefficients for energy eigenkets are, to first order

$$c_n(t) \approx c_n^{(0)}(t) + \lambda c_n^{(1)}(t)$$

$$= \delta_{ni} - \frac{i\lambda}{\hbar} \int_0^t V_{ni} e^{i\omega_{ni}t} dt$$

To first order, the transition probabilities $P(i \rightarrow n) = |c_n^{(1)}(t)|^2$. From the expression above, we have

$$c_n^{(1)}(t) = -\frac{i}{\hbar} \int_0^t V_{ni} e^{i\omega_{ni}t} dt$$

$$= \frac{V_{ni}}{E_n - E_i} (1 - e^{i\omega_{ni}t})$$

Therefore,

$$P(i \rightarrow n) = \frac{4|V_{ni}|^2}{|E_n - E_i|^2} \sin^2 \left(\frac{(E_n - E_i)t}{2\hbar} \right) \quad (1)$$

Clearly the transition probability oscillates but the amplitude of this oscillation is small provided the magnitude of the transition matrix element V_{ni} of the perturbation is small compared to the unperturbed energy difference between the initial and final states. However, if $E_n = E_i$, the transition probability grows quadratically in time

$$P(i \rightarrow n) = |V_{ni}|^2 t^2 \quad (2)$$

(I) To compute the wavefunction $\langle i|\alpha(\tau)\rangle$ we need both the magnitude and phase of the first order corrections to the expansion coefficients:

$$\begin{aligned} c_n(t) &\approx c_n^{(0)}(t) + \lambda c_n^{(1)}(t) \\ &= \delta_{ni} + \frac{V_{ni}}{E_n - E_i} (1 - e^{i\omega_{ni}t}) \end{aligned}$$

We can ensure that $\langle i|\alpha(\tau)\rangle$ is normalized by dividing the coefficients by $\sum_n |c_n(\tau)|^2$ at each τ .

(J) I wasn't able to get this part to work correctly with my FDTD method, so my animation is generated by computing the coefficients from part (I) directly and constructing the superposition:

$$\langle i|\alpha(\tau)\rangle = c_1(t)\psi_1(\ell) + c_2(t)\psi_2(\ell) + c_3(t)\psi_3(\ell)$$

where $\psi_n(\ell)$ are orthonormal.

(K) As $\tau \rightarrow \infty$ we should have the wavefunction approach that of the ground state i.e., $\langle i|\alpha(\tau)\rangle \rightarrow \langle i|1\rangle$, because as you can see in Figure 2, the transition probability $P(1 \rightarrow 1)$ grows quadratically in time while the other transition probabilities oscillate sinusoidally.

(L) The expectation value $\langle x/L \rangle$ is plotted as a function of time in Figure 3. When the potential is turned on, the particle goes from a stationary state to a superposition and then slowly back to a stationary state for the same reason as in (K).

(M) The expectation value $\langle H \rangle$ is plotted as a function of time in Figure 3.

(N) Energy does not need to be conserved by the uncertainty principle for energy and time $\Delta E \Delta t \geq \hbar/2$ which in this case is $\Delta E \Delta \tau \geq 1/2$. Since time is a parameter in nonrelativistic quantum mechanics, energy conservation can be violated over short time scales.

(O) At long times $\langle H_0 \rangle = E_0$, so no energy is transferred to the system

```

import numpy as np
import matplotlib.pyplot as plt
from numpy import linalg as LA
from plt2array import plt2array
from skimage.io import imsave

class FDTDSolver:
    def
        __init__(self, Nx, Nt, V, psi_r0, psi_i0, dir, plot=True, plot_iter_num=10, dt=0.1, name='sim', H=):
            self.Nt = Nt
            self.Nx = Nx
            self.V = V
            self.psi_r0 = psi_r0
            self.psi_i0 = psi_i0
            self.psi_r = psi_r0
            self.psi_i = psi_i0
            self.psi_r = np.pad(self.psi_r, (1,1))
            self.psi_i = np.pad(self.psi_i, (1,1))
            self.prob = np.zeros((Nx, Nt))
            self.prob[:,0] = self.psi_r[1:-1]**2 + self.psi_i[1:-1]**2
            self.c1 = dt
            self.c2 = dt
            self.dir = dir
            self.plot_iter_num = plot_iter_num
            self.plot = plot
            self.name = name
            self.dt = dt
            self.H = H
            self.Etau = np.zeros((Nt,))

    def compute_energy(self, t):
        psi = self.psi_r[1:-1] + self.psi_i[1:-1]*1j
        self.Etau[t] = np.conjugate(psi) @ self.H @ psi

    def plot_init_(self):
        fig, ax = plt.subplots(figsize=(3,1))
        ax.plot(self.psi_r0, color='red')
        ax.plot(self.psi_i0, color='blue')
        plt.tight_layout()
        plt.show()

    def plot_iter(self, t):
        fig, ax = plt.subplots()
        ax1 = ax.twinx()
        ax.set_ylim([-0.1, 0.1])
        ax1.set_ylim([-2*self.V.max(), 2*self.V.max()])
        ax.plot(self.psi_r, color='red', label=r'$\psi_{R}$')
        ax.plot(self.psi_i, color='blue', label=r'$\psi_{I}$')

```

```

ax.plot(self.psi_i**2 +
        self.psi_r**2,color='purple',label=r'$|\psi|^2$')
ax1.plot(self.V[:,t],color='black',label=r'$V(x)$')
ax1.set_ylabel('V(x)')
ax.text(0.05, 0.95, r'$\tau$' + f'={self.dt*t}', transform=ax.transAxes,
        fontsize=14,
        verticalalignment='top')
ax.legend()
plt.tight_layout()
rgb_array = plt2array(fig)
imsave(self.dir+f'{t}_{self.name}.tif',rgb_array)
plt.close()

def update_r(self,t):
    for n in range(1,self.Nx+1):
        self.psi_r[n] = self.psi_r[n]-\
            self.c1*(self.psi_i[n+1] - 2*self.psi_i[n] + self.psi_i[n-1]) +\
            self.c2*self.V[n,t]*self.psi_i[n]

def update_i(self,t):
    for n in range(1,self.Nx+1):
        self.psi_i[n] = self.psi_i[n] + \
            self.c1*(self.psi_r[n+1] - 2*self.psi_r[n] + self.psi_r[n-1])\
            - self.c2*self.V[n,t]*self.psi_r[n]
def forward(self):
    if self.plot:
        self.plot_iter(0)
    for t in range(1,self.Nt):
        if self.H is not None:
            self.compute_energy(t)
        print(f'Time step: {t}')
        self.update_i(t)
        self.update_r(t)
        self.prob[:,t] = self.psi_r[1:-1]**2 + self.psi_i[1:-1]**2
        if t % self.plot_iter_num == 0:
            if self.plot:
                self.plot_iter(t)

```

```

import numpy as np
import matplotlib.pyplot as plt
from numpy import linalg as LA
from fdtd import FDTDsolver

dir = '/home/cwseitz/Desktop/temp/'
Nx = 100
Nt = 400000
eta = 1
dtau = 0.01

```

```

#####
# Infinite square well
#####

print('Simulating the infinite square well...\n')
tau1 = np.arange(0,Nt)*dtau
V = np.zeros((Nx,Nt))
H = np.zeros((Nx,Nx)) #Hamiltonian at t = 0
H += np.diag(2*eta + V[:,0],k=0) #main diagonal
H += np.diag(-eta*np.ones((Nx-1,)),k=1) #upper diagonal
H += np.diag(-eta*np.ones((Nx-1,)),k=-1) #lower diagonal

#####
# Find eigenvectors and eigenvalues of H0
#####

vals, vecs = LA.eig(H)
idx = np.argsort(vals)
vecs1 = vecs[:,idx]
vals1 = vals[idx]

#####
# Simulate time evolution in a infinite square well
#####

V = np.pad(V, ((1,1),(0,0)))
psi_r0 = (vecs1[:,0] + vecs1[:,1])/np.sqrt(2)
psi_i0 = np.zeros_like(psi_r0)
solver1 =
    FDTDSolver(Nx,Nt,V,psi_r0,psi_i0,dir,plot_iter_num=10000,plot=True,dt=dtau,name='inf',H=H)
solver1.forward()

#####
# Finite square well
#####

Nt = 40000
print('Simulating the finite square well...\n')
tau2 = np.arange(0,Nt)*dtau
Vl = 2
Vr = 3
V = np.zeros((Nx,Nt))
V[:30,:] = Vl
V[70:,:] = Vr
H = np.zeros((Nx,Nx)) #Hamiltonian at t = 0
H += np.diag(2*eta + V[:,0],k=0) #main diagonal
H += np.diag(-eta*np.ones((Nx-1,)),k=1) #upper diagonal
H += np.diag(-eta*np.ones((Nx-1,)),k=-1) #lower diagonal

```



```
#####
# Find eigenvectors and eigenvalues of H0
#####

vals, vecs = LA.eig(H)
idx = np.argsort(vals)
vecs2 = vecs[:,idx]
vals2 = vals[idx]

#####
# Simulate time evolution in a finite square well
#####

V = np.pad(V, ((1,1),(0,0)))
psi_r0 = (vecs2[:,0] + vecs2[:,1])/np.sqrt(2)
psi_i0 = np.zeros_like(psi_r0)
solver2 =
    FDTDsolver(Nx,Nt,V,psi_r0,psi_i0,dir,plot_iter_num=1000,plot=True,dt=dtau,name='fin',H=H)
solver2.forward()

#####
# Plot expectation values of scaled position
#####

fig, ax = plt.subplots(1,3,figsize=(6,2))
X_avg = solver1.prob.T * np.arange(0,Nx)
X_avg = np.sum(X_avg,axis=1)
ax[0].plot(tau1,X_avg,color='black')
ax[0].set_xlabel(r'$\tau$')
ax[0].set_ylabel(r'$\langle x/L \rangle$')
ax[0].set_ylim([30,70])
X_avg = solver2.prob.T * np.arange(0,Nx)
X_avg = np.sum(X_avg,axis=1)
ax[1].plot(tau2,X_avg,color='purple')
ax[1].set_xlabel(r'$\tau$')
ax[1].set_ylabel(r'$\langle x/L \rangle$')
ax[1].set_ylim([30,70])
ax[2].plot(vals1,color='red',label='ISW')
ax[2].plot(vals2,color='blue',label='FSW')
ax[2].set_label('n')
ax[2].set_ylabel(r'$\epsilon_n$')
ax[2].legend(fontsize=8)
plt.tight_layout()
plt.show()
```

```
import numpy as np
import matplotlib.pyplot as plt
```

```

from numpy import linalg as LA
from fdtd import FDTDSolver

dir = '/home/cwseitz/Desktop/temp/'
Nx = 100
Nt = 50000
eta = 1 #hopping parameter
dtau = 0.1

#####
# Infinite square well
#####

V = np.zeros((Nx,Nt))
H = np.zeros((Nx,Nx)) #Hamiltonian at t = 0
H += np.diag(2*eta + V[:,0],k=0) #main diagonal
H += np.diag(-eta*np.ones((Nx-1,)),k=1) #upper diagonal
H += np.diag(-eta*np.ones((Nx-1,)),k=-1) #lower diagonal

#####
# Find eigenvectors and eigenvalues of H0
#####

vals, vecs = LA.eig(H)
idx = np.argsort(vals)
vecs1 = vecs[:,idx]
vals1 = vals[idx]

#####
# Define the perturbation Hamiltonian
#####

Vl = 2
Vr = 3
V = np.zeros((Nx,Nt))
V[:30,:] = Vl
V[70:,:] = Vr
lam = 5 * 10**-4
V *= lam

#####
# Transform the perturbation to energy basis
#####

Hp = np.zeros((Nx,Nx)) #perturbation Hamiltonian
Hp += np.diag(V[:,0],k=0) #main diagonal
U0 = LA.inv(vecs1) #unitary operator
Hp_e = U0 @ Hp @ LA.inv(U0) #perturbation in energy basis
P_0n = Hp_e[0,1:]**2

```

```

E0 = vals1[0]
P_On /= (vals1[1:]-E0)**2

#####
# Compute transition probability for n=1,2,3
#####

tau = np.arange(0,Nt)*dtau
v_11 = Hp_e[0,0]
v_12 = Hp_e[0,1]
v_13 = Hp_e[0,2]
c_11_msq = (tau*v_11)**2
c_12_msq =
    ((4*v_12**2)/((vals1[1]-vals1[0]))**2)*np.sin(tau*(vals1[1]-vals1[0])/2)**2
c_13_msq =
    ((4*v_13**2)/((vals1[2]-vals1[0]))**2)*np.sin(tau*(vals1[2]-vals1[0])/2)**2
Z = c_11_msq + c_12_msq + c_13_msq
c_11_msq /= Z
c_12_msq /= Z
c_13_msq /= Z
Ebar = c_11_msq*vals1[0] + c_12_msq*vals1[1] + c_13_msq*vals1[2]

fig, ax = plt.subplots(1,3,figsize=(7,2))
ax[0].plot(tau,c_11_msq,color='red',label=r'$|c_{11}|^2$')
ax[1].plot(tau,c_12_msq,color='blue',label=r'$|c_{12}|^2$')
ax[2].plot(tau,c_13_msq,color='purple',label=r'$|c_{13}|^2$')
ax[0].set_xlabel(r'$\tau$')
ax[1].set_xlabel(r'$\tau$')
ax[2].set_xlabel(r'$\tau$')
ax[0].set_ylabel(r'$P(1 \rightarrow 1)$')
ax[1].set_ylabel(r'$P(1 \rightarrow 2)$')
ax[2].set_ylabel(r'$P(1 \rightarrow 3)$')
plt.tight_layout()
plt.show()

#####
# Simulate time evolution for a time dependent Hamiltonian
#####

V = np.pad(V, ((1,1),(0,0)))
psi_r0 = -1*vecs1[:,0] #pure ground state
psi_i0 = np.zeros_like(psi_r0)
solver =
    FDTDsolver(Nx,Nt,V,psi_r0,psi_i0,dir,plot_iter_num=5000,plot=True,dt=dtau,H=H+Hp)
solver.forward()

#####

```

```

# Construct time evolution analytically
#####

probt = np.zeros((Nx,Nt))
probt[:,0] = vecs1[:,0]**2
print(np.sum(probt[:,0]))
for t in range(1,Nt):
    psi = c_11_msq[t]*vecs1[:,0] + c_12_msq[t]*vecs1[:,1] + c_13_msq[t]*vecs1[:,2]
    probt[:,t] = (psi**2)/np.sum(psi**2)
p1 = probt[:,1]*np.arange(0,Nx)
X_avg = probt.T * np.arange(0,Nx)
X_avg = np.sum(X_avg,axis=1)
X_avg[0] = 50
fig, ax = plt.subplots(1,2)
ax[0].plot(tau,X_avg[0:],color='black')
ax[0].set_xlabel(r'$\tau$')
ax[0].set_ylabel(r'$\langle x/L \rangle$')
ax[1].plot(tau,Ebar,color='black')
ax[1].set_xlabel(r'$\tau$')
ax[1].set_ylabel(r'$\langle H \rangle$')
plt.tight_layout()
plt.show()

#####
# Position/Energy expectation value for time-dependent Hamiltonian
#####

X_avg = solver.probt.T * np.arange(0,Nx)
X_avg = np.sum(X_avg,axis=1)
tau = np.arange(0,Nt)*dttau
fig, ax = plt.subplots()
ax.plot(tau,X_avg,color='black')
ax.set_xlabel(r'$\tau$')
ax.set_ylabel(r'$\langle x/L \rangle$')
plt.tight_layout()
plt.show()

```
