Lecture9

February 28, 2019

1 Lecture 9: Shooting, Matching and finite difference

Overview: * Time independent Schrödinger eqn. * Numerov's method. * Finite difference formula (discrete Laplacian). * Displacement eqn.

Next Lecture: * Boundary value problems, Laplace and Poisson equations. —

1.1 Tasks

1.1.1 Shooting:

- How does the number of bound states in the square well change when you change the depth
 of the well? What about changing the width of the well? Do your finding agree with your
 expectations?
- Is it possible a minimum well depth above which there is no bound-state?

1.1.2 Finite Difference

- Modify the code to include a simple harmonic potential i.e $V = 1/2kx^2$.
- Compare the accuracy of the matching and finite difference methods for a simple harmonic potential. Which method do you expect to be more accurate? How do your results compare with the exact eigen energies?
- Try defining some other interesting potentials e.g. double square well.

```
self.xarray = np.linspace(x_start, x_end, npoints, endpoint = True) # include
             def scipy_trajectory(self):
                 """calculate trajectory using SciPy ode integrator"""
                 self.uv = odeint(self.sch, self.uv0, self.xarray)
                 self.uv_end = self.uv[-1]
             def sch(self, uv, x):
                 """right hand side of the differential equation"""
                 u = uv[1]
                 v = 2*(self.Potential(x) - self.E)*uv[0]
                 return np.ravel(np.array([u, v]))
             def Potential(self, x):
                 # finite square well potential
                 a = self.a # width of well
                 VO = self.VO # depth of well
                 if (abs(x) > a/2.):
                     return 0
                 else :
                     return -V0
In [15]: # the matching function
         def match(En):
             \#a = 1.0e-11
             p_up = QM(E = En, npoints = 1000, x_start = -10)
             p_down = QM(E = En, npoints = 1000, x_start = 10)
             p_up.scipy_trajectory()
             p_down.scipy_trajectory()
             return p_down.uv_end[0]*p_up.uv_end[1] - p_down.uv_end[1]*p_up.uv_end[0]
In [27]: a = 1.0e-3
         E1 = -10 \# start Energy search at the well depth
         dE = 0.01
         npoints = 1000
         E_SM = []
        fig = plt.figure()
         ax = fig.add_subplot(111)
         # Bound states must have E < 0 for the square well.
         # Cut off after we have found a few states
```

always want to start at furthest point and go to zero

```
while (E1 < -8-dE):
             if match(E1)*match(E1 + dE) < 0 : # bracket E</pre>
                 # find the energy
                 E = rtf.bisect(match, E1, E1+dE, 1e-12)
                 print('Energy found: %.5f'%(E))
                 E SM.append(E)
                 dx = 16*a/npoints
                 # for plotting, must ensure overlap of left and right solutions
                 p_up = QM(E, npoints = npoints + 4, x_start = -10, x_end = 0 + 4 * dx)
                 p_down = QM(E, npoints = npoints-4, x_start = 10, x_end = 0+4*dx)
                 p_up.scipy_trajectory()
                 p_down.scipy_trajectory()
                 # scale factor
                 scale = p_up.uv_end[0]/p_down.uv_end[0]
                 # full solution, combine up and down solutions
                 psi_x = np.concatenate((p_up.uv[:-1,0], scale*p_down.uv[::-1,0]))
                 xa = np.linspace(-10,10, 2*npoints-1, endpoint = True)
                 # plot the scaled solution (not normalized), scale maximum to 1 for plotting
                 ax.plot(xa, psi_x/max(psi_x), lw = 2, label = "E = {:.3f}".format(E))
                  print("No bracket, increasing energy to", E1+dE)
             E1 += dE
         # Solution is exact only for for wide and deep well, and odd n
         a = 5
         VO = 10
         n = np.arange(1,5,2)
         Exact = -V0+np.pi**2/2/a**2*n**2
         print(Exact)
         # plot the potential
         ax.plot(xa, np.vectorize(p_up.Potential)(xa)/10., color = 'k', lw =2)
         ax.set_xlim([-8, 8])
         ax.set_xlabel('x', fontsize = 14)
         ax.set_ylabel('$\psi$ (unnormalized)', fontsize = 14)
         ax.legend(loc = 4, fontsize =10, numpoints = 1 )
         plt.show()
<IPython.core.display.Javascript object>
<IPython.core.display.HTML object>
```

```
Energy found: -9.83377
Energy found: -9.33600
Energy found: -8.50973
[-9.80260791 -8.22347121]
In [41]: def match(En):
             \#a = 1.0e-11
             p_up = QM(E = En, npoints = 1000, x_start = -10, a = 6)
             p_down = QM(E = En, npoints = 1000, x_start = 10, a = 6)
             p_up.scipy_trajectory()
             p_down.scipy_trajectory()
             return p_down.uv_end[0]*p_up.uv_end[1] - p_down.uv_end[1]*p_up.uv_end[0]
         a = 1.0e-3
         E1 = -10 \# start Energy search at the well depth
         dE = 0.01
         npoints = 1000
         E_SM = []
         fig2 = plt.figure()
         ax2 = fig2.add_subplot(111)
         # Bound states must have E < 0 for the square well.
         # Cut off after we have found a few states
         while (E1 < -8-dE):
             if match(E1)*match(E1 + dE) < 0 : # bracket E</pre>
                 # find the energy
                 E = rtf.bisect(match, E1, E1+dE, 1e-12)
                 print('Energy found: %.5f'%(E))
                 E_SM.append(E)
                 dx = 16*a/npoints
                 # for plotting, must ensure overlap of left and right solutions
                 p_up = QM(E, npoints = npoints + 4, x_start = -10, x_end = 0 + 4 * dx, a = 6)
                 p_down = QM(E, npoints = npoints-4, x_start = 10, x_end = 0+4*dx, a =6)
                 p_up.scipy_trajectory()
                 p_down.scipy_trajectory()
```

```
# scale factor
                 scale = p_up.uv_end[0]/p_down.uv_end[0]
                 # full solution, combine up and down solutions
                 psi_x = np.concatenate((p_up.uv[:-1,0], scale*p_down.uv[::-1,0]))
                 xa = np.linspace(-10,10, 2*npoints-1, endpoint = True)
                 # plot the scaled solution (not normalized), scale maximum to 1 for plotting
                 ax2.plot(xa, psi_x/max(psi_x), lw = 2, label = "E = {:.3f}".format(E))
             #else:
                  print("No bracket, increasing energy to", E1+dE)
             E1 += dE
         ax2.plot(xa, np.vectorize(p_up.Potential)(xa)/10., color = 'k', lw =2)
         ax2.set_xlim([-8, 8])
         ax2.set_xlabel('x', fontsize = 14)
         ax2.set_ylabel('$\psi$ (unnormalized)', fontsize = 14)
         ax2.legend(loc = 4, fontsize =10,numpoints = 1 )
         plt.show()
<IPython.core.display.Javascript object>
<IPython.core.display.HTML object>
Energy found: -9.88131
Energy found: -9.52565
Energy found: -8.93428
Energy found: -8.10954
In [45]: def match(En):
             \#a = 1.0e-11
             p_up = QM(E = En, npoints = 1000, x_start = -10, V0 = 8.5)
             p_down = QM(E = En, npoints = 1000, x_start = 10, V0 = 8.5)
             p_up.scipy_trajectory()
             p_down.scipy_trajectory()
             return p_down.uv_end[0]*p_up.uv_end[1] - p_down.uv_end[1]*p_up.uv_end[0]
         a = 1.0e-3
         E1 = -10 \# start Energy search at the well depth
         dE = 0.01
         npoints = 1000
```

```
fig2 = plt.figure()
         ax2 = fig2.add_subplot(111)
         # Bound states must have E < 0 for the square well.
         # Cut off after we have found a few states
         while (E1 < -8-dE):
             if match(E1)*match(E1 + dE) < 0 : # bracket E</pre>
                 # find the energy
                 E = rtf.bisect(match, E1, E1+dE, 1e-12)
                 print('Energy found: %.5f'%(E))
                 E_SM.append(E)
                 dx = 16*a/npoints
                 # for plotting, must ensure overlap of left and right solutions
                 p_up = QM(E, npoints = npoints+4, x_start = -10, x_end = 0+4*dx, VO = 8.5)
                 p_down = QM(E, npoints = npoints-4, x_start = 10, x_end = 0+4*dx, VO = 8.5)
                 p_up.scipy_trajectory()
                 p_down.scipy_trajectory()
                 # scale factor
                 scale = p_up.uv_end[0]/p_down.uv_end[0]
                 # full solution, combine up and down solutions
                 psi_x = np.concatenate((p_up.uv[:-1,0], scale*p_down.uv[::-1,0]))
                 xa = np.linspace(-10,10, 2*npoints-1, endpoint = True)
                 # plot the scaled solution (not normalized), scale maximum to 1 for plotting
                 ax2.plot(xa, psi_x/max(psi_x), lw = 2, label = "E = {:.3f}".format(E))
             #else :
                  print("No bracket, increasing energy to", E1+dE)
             E1 += dE
         ax2.plot(xa, np.vectorize(p_up.Potential)(xa)/10., color = 'k', lw =2)
         ax2.set_xlim([-8, 8])
         ax2.set_xlabel('x', fontsize = 14)
         ax2.set_ylabel('$\psi$ (unnormalized)', fontsize = 14)
         ax2.legend(loc = 4, fontsize =10, numpoints = 1 )
         plt.show()
<IPython.core.display.Javascript object>
<IPython.core.display.HTML object>
```

 $E_SM = []$

1.2 Finite difference

```
In [17]: def potential(x):
             """ The potential energy for a finite square well"""
            a = 5
            VO = 10
            V = np.zeros(x.shape)
            V[np.nonzero(np.abs(x)<a/2)] = -V0
            return V
In [18]: N = 2000 # number of lattice points including boundaries at 0, N-1
        L = 100.0
        dx = L / N # x runs from -L/2 + dx/2 to L/2 + dx/2
        \# D = array(zeros((N, N))) \# discrete laplacian operator
        D = np.diag([-2]*(N))
        D+= np.diag([1]*(N-1),1) + np.diag([1]*(N-1),-1)
        print("\nLattice Laplacian operator")
        print(D)
        x = np.linspace(0.5*(dx-L), 0.5*(L-dx), N)
        V = np.diag(potential(x)) # potential (the potential is a diagonal matrix because it'
        H = -0.5 * pow(dx, -2.0) * D + V # Hamiltonian. Here m = hbar = 1
        print("\nMatrix elements of Hamiltionian = ")
        print(H)
Lattice Laplacian operator
[[-2 1 0 ... 0 0 0]
[ 1 -2 1 ... 0 0 0]
 [ 0 1 -2 ... 0 0 0]
 [ 0 0 0 ... -2 1 0]
 [ 0 0 0 ... 1 -2 1]
 [0 \ 0 \ 0 \dots \ 0 \ 1 \ -2]]
Matrix elements of Hamiltionian =
[[ 400. -200. 0. ...
                          0.
                                0.
                                      0.]
 [-200. 400. -200. ...
                                0.
                                      0.1
                          0.
    0. -200. 400. ...
                          0.
                                0.
                                      0.1
    0.
          0.
                0. ... 400. -200.
    0.
          0.
                0. ... -200. 400. -200.]
```

```
[ 0. 0. 0. ... 0. -200. 400.]]
In [19]: E_FD, v = eigh(H) # diagonalize Hamiltonian
         # eigh returns the array of shape (N,N) where each column is an eigenvector;
         # we transpose for plotting later
        v = v.transpose()
        print("\nGround state energy = ", E_FD[0])
        print("Energies of low-lying excited states = ", E_FD[1], E_FD[2], E_FD[3], E_FD[4])
Ground state energy = -9.83394417552133
Energies of low-lying excited states = -9.336822780606123 -8.511972729345871 -7.3657397173162
In [20]: fig = plt.figure()
        ax = fig.add_subplot(111)
         #plot potential (scaled)
        ax.plot(x, potential(x)/np.max(np.abs(potential(x))), lw = 2, color = 'grey')
        ax.plot(x, v[0]/np.max(v[0]), lw =2, label = "E = {:.3f}".format(E_FD[0]))
         ax.plot(x, v[1]/np.max(v[1]), lw = 2, label = "E = {:.3f}".format(E_FD[1]))
         ax.plot(x, v[2]/np.max(v[2]), lw = 2, label = "E = {:.3f}".format(E_FD[2]))
         ax.set_xlabel("x", fontsize = 14)
        ax.set_ylabel('$\psi$ (unnormalized)', fontsize = 14)
        ax.set_xlim([-8,8])
         \#ax.set\_ylim([-1, 1])
         ax.legend(loc = 4, fontsize =10, numpoints = 1)
<IPython.core.display.Javascript object>
<IPython.core.display.HTML object>
Out[20]: <matplotlib.legend.Legend at 0x12f9564a8>
In [9]: # Comparision of Shooting and Matching with Finite Difference:
        print("\nGround state energy")
        print("Matching: \t\t", E_SM[0], "\nFinite Difference:\t", E_FD[0])
       print("\nFirst excite state energy")
        print("Matching: \t\t", E_SM[1], "\nFinite Difference:\t", E_FD[1])
        print("\n2nd excited state energy")
        print("Matching: \t\t", E_SM[2], "\nFinite Difference:\t", E_FD[2])
```

Ground state energy

Matching: -9.833765216465958 Finite Difference: -9.83394417552133

First excite state energy

Matching: -9.336000745337868 Finite Difference: -9.336822780606123

2nd excited state energy

Matching: -8.50973393157125 Finite Difference: -8.511972729345871

In []: