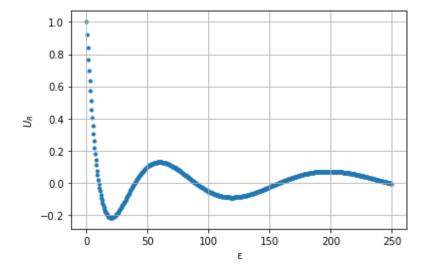
```
In [3]: | import numpy as np
        import matplotlib.pyplot as plt
        from scipy.stats import linregress
        import pandas as pd
        # 2(a)
        def RK4(func, X0, tmin, tmax, N):
            h = (tmax-tmin)/N
            t = np.linspace(tmin, tmax, N+1)
            X = np.zeros([N+1, len(X0)])
            X[0] = X0
            for i in range(N):
                k1 = func(t[i], X[i])
                k2 = func(t[i] + h/2, X[i] + (h * k1)/2)
                k3 = func(t[i] + h/2, X[i] + h/2 * k2)
                k4 = func(t[i] + h, X[i] + h * k3)
                X[i+1] = X[i] + h / 6. * (k1 + 2*k2 + 2*k3 + k4)
            return X, t
        def simps(x, y):
            h = (x[-1]-x[0])/len(x)
            integral = (h/3)*(2*np.sum(y[2:-2:2]) + 4*np.sum(y[1:-1:2]) + y[0] + y
        [-1])
            return integral
        # 2(a) i.----
        e = np.arange(0, 250, 0.5)
        index, index_, u_r = [], [], []
        for i in e:
            def func(x, Y):
                y, y1 = Y # y1 is the first order derivative
               f1 = y1
                f2 = -(i)*y
                return np.array([f1, f2])
            x, t = RK4(func, [0, 1], -1/2, 1/2, 100)
            u = x[:, 0][-1]
            u_r.append(u)
        for i in range(1, len(e)):
            if u_r[i-1]*u_r[i] < 0:</pre>
                index.append(i-1)
                index_.append(i)
        plt.scatter(e, u_r, s=10)
        plt.xlabel('ε')
        plt.ylabel(r'$U_R$')
```

```
plt.grid()
plt.show()
# 2(a) ii.
print('indices of energy vector:', index)
print('the corresponding values of energies:', [e[i] for i in index])
# 2(a) iii.
sec_e = []
for i, j in zip(index, index_):
    # guessing the value of 'e' using secant method
    approx_e = e[j] - u_r[j]*((e[j]-e[i])/(u_r[j]-u_r[i]))
    sec_e.append(approx_e)
print('the final energy eigen values:', sec_e)
n, j = 1, 2
for i in sec_e:
   def func(x, Y):
        y, y1 = Y # y1 is the first order derivative
       f1 = y1
        f2 = -(i)*y
        return np.array([f1, f2])
    if n % 2 == 0: # the func is even
        if n == 2:
            inc = -1
        else:
            inc = 1
        x, t = RK4(func, [0, inc], -1/2, 1/2, 100)
        wfn = x[:, 0] # the wavefunction
        # the normalized wave function
        nwfn = wfn/(np.sqrt(simps(t, wfn**2)))
        ana = np.sin(np.pi*t*n) # analytical solution
        # normalised analytical solution
        n_ana = ana/(np.sqrt(simps(t, ana**2)))
    else: # the func is odd
        if n == 3:
            inc = -1
        else:
            inc = 1
        x, t = RK4(func, [0, inc], -1/2, 1/2, 100)
        wfn = x[:, 0] # the wavefunction
        # the normalized wave function
        nwfn = wfn/(np.sqrt(simps(t, wfn**2)))
        ana = np.cos(np.pi*n*t) # analytical solution
        # normalized analytical solution
        n_ana = ana/(np.sqrt(simps(t, ana**2)))
```

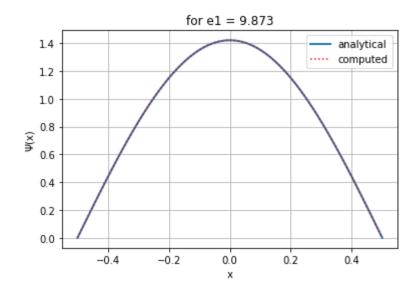
```
plt.plot(t, n_ana, label='analytical', linewidth='2')
   plt.plot(t, nwfn, label='computed', ls='dotted', color='red')
   plt.xlabel('x')
   plt.ylabel('\Psi(x)')
   plt.title(f'for e{n} = {round(i,3)}')
   plt.grid()
   plt.legend()
   plt.show()
   j += 1
   n += 1
# 2(b) -----
n = np.array([1, 2, 3, 4, 5])
slope = linregress(n**2, sec_e)[0]
intercept = linregress(n**2, sec_e)[1]
sec_ey = slope*(n**2) + intercept # fitted points
plt.scatter(n**2, sec_e)
plt.plot(n**2, sec_ey)
plt.xlabel(r'$n^2$')
plt.ylabel(r'$e_n$')
plt.title(r'$e_n$ as a function of $n^2$')
plt.grid()
plt.show()
print('slope:', slope)
# 2(c) -----
n, j = 1, 8
for i in sec_e:
   def func(x, Y):
       y, y1 = Y # y1 is the first order derivative
       f1 = y1
       f2 = -(i)*y
       return np.array([f1, f2])
   if n % 2 == 0: # the func is even
       if n == 2:
           inc = -1
       else:
           inc = 1
       x, t = RK4(func, [0, inc], -1/2, 1/2, 100)
       wfn = x[:, 0] # the wavefunction
       # the normalized wave function
       nwfn = wfn/(np.sqrt(simps(t, wfn**2)))
       ana = np.sin(np.pi*t*n) # analytical solution
       # normalised analytical solution
       n_ana = ana/(np.sqrt(simps(t, ana**2)))
```

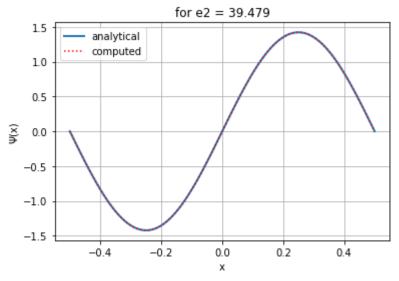
```
else: # the func is odd
       if n == 3:
           inc = -1
       else:
           inc = 1
       x, t = RK4(func, [0, inc], -1/2, 1/2, 100)
       wfn = x[:, 0] # the wavefunction
       # the normalized wave function
       nwfn = wfn/(np.sqrt(simps(t, wfn**2)))
       ana = np.cos(np.pi*n*t) # analytical solution
       # normalized analytical solution
       n_ana = ana/(np.sqrt(simps(t, ana**2)))
   plt.plot(t, n_ana**2, label='analytical', linewidth='2')
   plt.plot(t, nwfn**2, label='computed', ls='dotted', color='red')
   plt.xlabel('x')
   plt.ylabel(r'$\Psi(x)^2$')
   plt.title(f'probability density plot for e{n} = {round(i,3)}')
   plt.grid()
   plt.legend()
   plt.show()
   j += 1
   n += 1
# 2(d) -----
n = np.array([1, 2, 3, 4, 5])
e = 1.6*(10**(-19)) # charge on an electron
h_{cut} = 1.054571817*(10**(-34)) # in Js
m_e = 9.1*(10**(-31)) # mass of electron in kgs
m_p = 1.67262192 * (10**(-27)) # mass of proton in kgs
L1 = 5*(10**(-10)) # width of the well in meters (angstrom to m)
L2 = 10*(10**(-10))
L3 = 5*(10**(-15)) # fermi meter to m
sec_e = np.array(sec_e)
def table(m, L):
   # divided by e for converting into eVs
   cal_E = sec_e*(h_cut**2)/(e*2*m*(L**2))
   ana_e = (n**2)*(np.pi**2)*(h_cut**2)/(e*2*m*(L**2))
   table_ = pd.DataFrame(
       {'computed E': np.round(cal_E, 6), 'analytical E': np.round(ana_e,
6)})
   return table_
print('energies in eV of an electron trapped in a well of 5 Å')
print(table(m_e, L1))
```

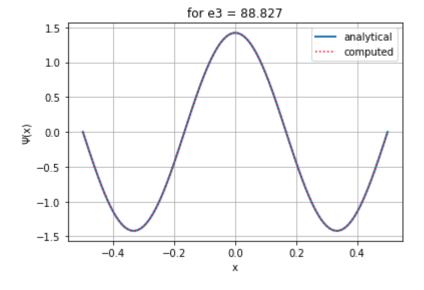
```
# 2(e)-----
print('energies in eV of an electron trapped in a well of 10 Å')
print(table(m_e, L2))
print('energies in eV of a proton trapped in a well of 5 fm')
print(table(m_p, L3))
# 2(f) -----
def funcn(x, Y):
   y, y1 = Y
   f1 = y1  # y1 is the first order derivative
   f2 = -(1)*y # e=1 for ground state
   return np.array([f1, f2])
x, t = RK4(funcn, [0, 1], -1/2, 1/2, 100)
u = x[:, 0] # the wavefunction
u_ = x[:, 1] # derivative of the wavefunction
exp_x = simps(t, (u**2)*t) # expectation value of x
\exp_x 2 = \operatorname{simps}(t, (u^{**}2)^*(t^{**}2)) # expectation value of x^2
exp_p = simps(t, u*u_) # expectation value of momentum
exp_p2 = simps(t, u*(u_**2)) # expectation value of momentum^2
var_x = exp_x2 - (exp_x**2)
var_p = exp_p2 - (exp_p**2)
# verifying the uncertainity principle
print('\Delta x \Delta p = ', var_x*var_p)
print('h/4\pi =', h_cut/2)
```

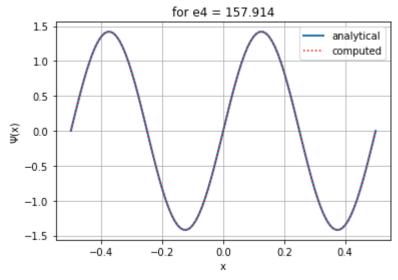


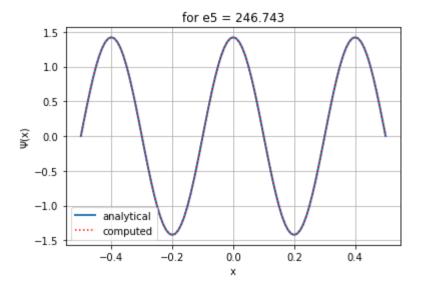
indices of energy vector: [19, 78, 177, 315, 493] the corresponding values of energies: [9.5, 39.0, 88.5, 157.5, 246.5] the final energy eigen values: [9.873225730254942, 39.47861914121265, 88.82 703027446989, 157.91448858562188, 246.7427816632612]



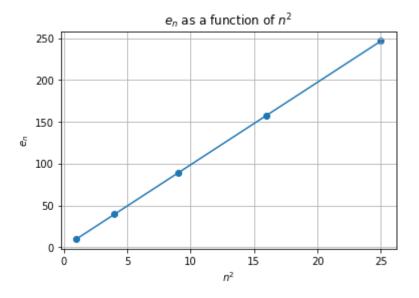




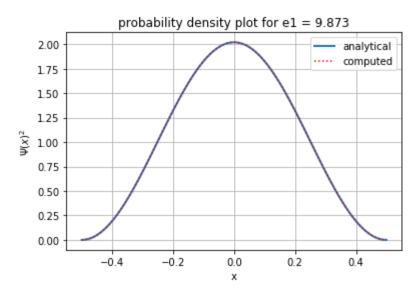


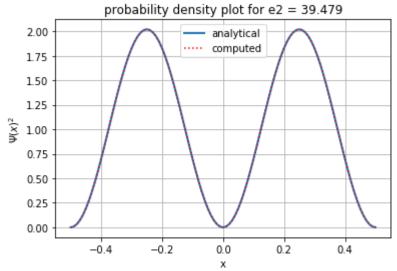


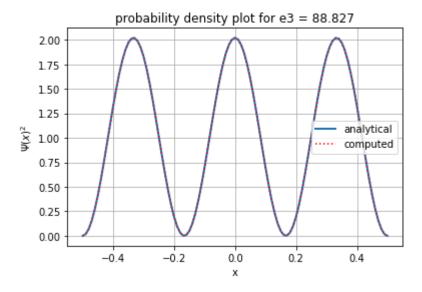
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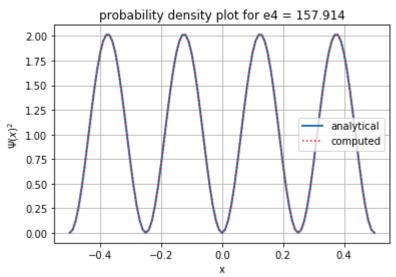


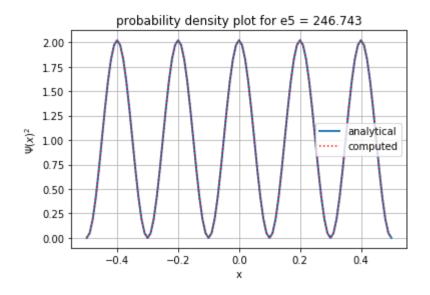
slope: 9.86961158923473











```
energies in eV of an electron trapped in a well of 5 Å
  computed E analytical E
    1.508273
                  1.507720
0
    6.030911
                  6.030880
1
2 13.569570
                 13.569480
3
   24.123645
                 24.123520
   37.693407
                 37.692999
energies in eV of an electron trapped in a well of 10 Å
  computed E analytical E
    0.377068
                   0.37693
0
1
    1.507728
                   1.50772
2
    3.392392
                   3.39237
3
    6.030911
                   6.03088
4
    9.423352
                   9.42325
energies in eV of a proton trapped in a well of 5 fm
    computed E analytical E
0 8.205851e+06 8.202841e+06
1 3.281153e+07 3.281136e+07
2 7.382606e+07 7.382557e+07
3 1.312461e+08 1.312455e+08
4 2.050732e+08 2.050710e+08
\Delta x \Delta p = 0.0033081688651695742
h/4\pi = 5.272859085e-35
```

In [ ]: