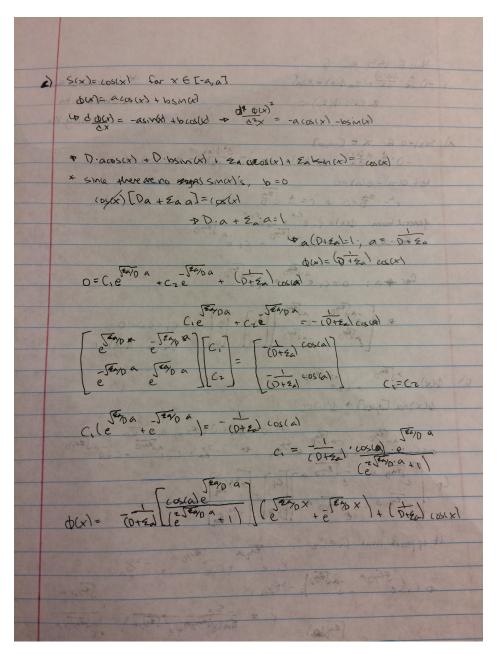
Assignment 6

)	
	Nuc. E 155 Hw 6
1)	-D & do(x) + Za b(x) = S(x) * (worked in this)
	C. C. : Q(ta)=0 problem with
,	anned Company - x 5 - Shand Book - Shakemity Vi
a.l	S(x)=0 for x E [-a,a] -D. & d (x) = 2 d (x) +P D (x) - 2 a d (x) = 0
	let (x-a)=0(a)=0 12-26=0 + (= +)26 12-26=0 + (= +)26
	12- 10=0 + (= 1 V D)
	general form: O(x)= (1 e x + cz e x x = x = x = x = x = x = x = x = x =
	for-a: 0= (, ea) 0 + (ze
	Com Cary
	far +a; 0=0, e + Cze
	p e e e (c, = cz = 0
	= (= (= 0) (= ()) (= ()) (= () = 0) (= 0) () (= () = 0) () () () () () () () () ()
	Ø(x)= c, e + (2e)
6)	Q(x)=C,e + cze
	0=C, a [[] + 0 (x)=0 0=C, a [] + cre 0=C, e + cre 1
	Fezo - 1 Exp 0+ 2a (a) = So, a = Eu - 20/0
	0=C, e + Cze + Za + - Za = C, e + Cze
	[Fly a - Jens a] [50/]
	(p [e = 50% a] [c1] = 50/24 50/24 50/24
	- 1200 a JENO a [Cr] 50/20
	It appears that Ci=Cz Fegoa - Jegoa - so
	Cie + Cie = Za
	DC.(e +e = -50/2a
	C1 = 5. (Jana - Jana) Janoa
	52% (21) 20/E 10 + E
-	- 500
	It appears that $C_1 = C_2$ $C_1 = C_1 = C_2$ $C_1 = C_1 = C_2$ $C_2 = C_1 = C_2$ $C_1 = C_2 = C_2$ $C_2 = C_2$ $C_3 = C_4$ $C_4 = C_4$ $C_4 = C_4$ $C_5 = C_4$ $C_4 = C_4$ $C_5 = C_4$ $C_5 = C_4$ $C_6 = C_4$ $C_7 = C_8$ $C_8 = C_8$ $C_8 = C_8$ $C_9 = C_9$ $C_$
	D(x)= -Soe (x) [x) [x]
	Ea(e wegoti) Le



2)

Code: $NE155_hw6_2.py$ import numpy as np
import matplotlib.pyplot as plt a = 4 #values are in cm D = 1Sigma = 0.2 # 1/cm

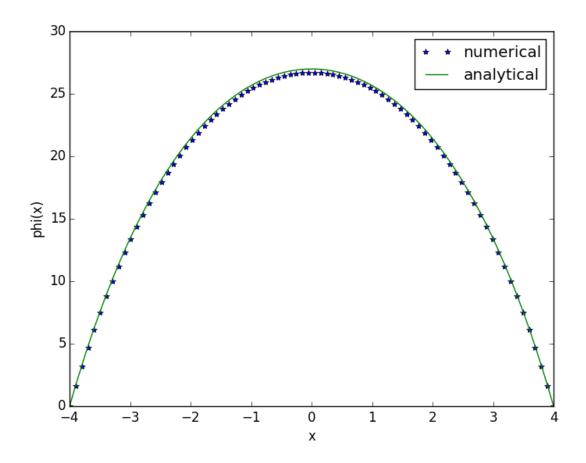
h = 0.1 #cm

L = (D/Sigma)**0.5

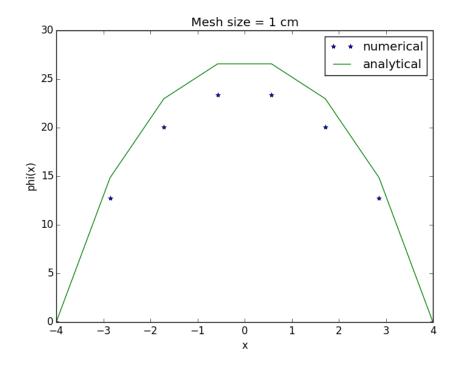
 $S = 8 \# n/(cm^3*s)$

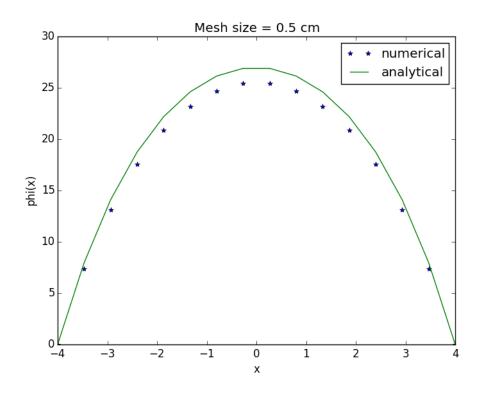
```
b array = np.zeros((78, 1))
 i = 0
 while i < 78:
         b array[i][0] = 8
         i += 1
 A = -D/(h ** 2)
 B = D * (2 + (h^{**}2)/(L^{**}2))/(h^{**}2)
 C = -D/(h^{**}2)
 def matrix a():
         a = []
         c = []
         i = 0
         while i < 77:
                 a.append(A)
                 c.append(C)
                 i += 1
         b = []
         i = 0
         while i < 78:
                 b.append(B)
                 i += 1
         def tridiag(a, b, c, k1=-1, k2=0, k3=1):
                  return np.diag(a, k1) + np.diag(b, k2) + np.diag(c, k3)
         M = tridiag(a, b, c)
         return M
 Am = matrix a()
 print(Am)
sol = np.dot(np.linalg.inv(Am), b array)
 print(sol)
 new sol = []
 new sol.append(0)
 for i in sol:
         new_sol.append(i)
new sol.append(0)
 x axis = np.linspace(-4, 4, 80)
 phi val = []
 for i in x_axis:
 8*np.exp(a*(Sigma/D)**0.5))/(Sigma*(np.exp(2*a*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1))*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Sigma/D)**0.5)+1)*(np.exp(i*(Si
 +np.exp(-i*(Sigma/D)**0.5)) + 8/Sigma
         phi val.append(phi)
plt.plot(x axis, new sol, '*')
plt.plot(x axis, phi val)
```

```
plt.xlabel('x')
plt.ylabel('phi(x)')
plt.legend(('numerical', 'analytical'))
plt.show()
```



3)





10

-3

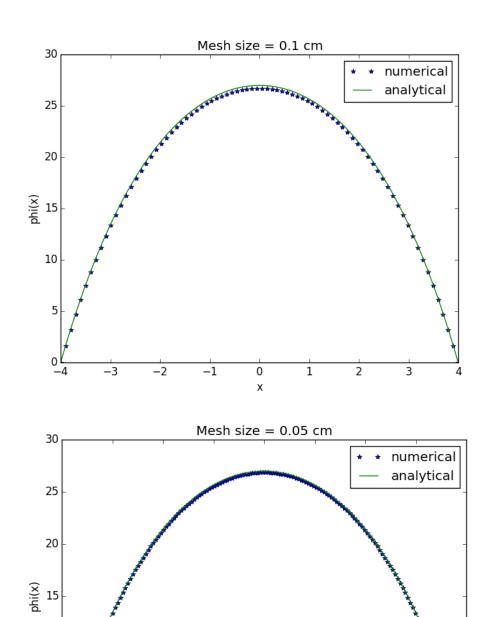
-2

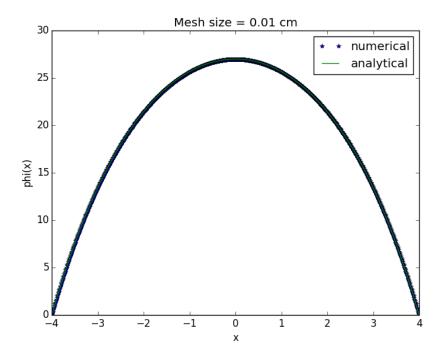
-1

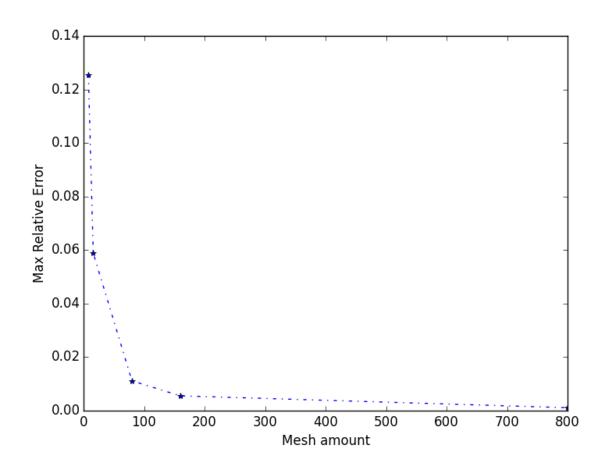
0 x 3

1

2







The higher the mess amount, the less the max relative error becomes. The reason for this is because decreasing the mesh size increases the number of calculation which in turn increases the accuracy. We can also see that there is an exponential trend between the mesh amount and the relative error. This is because the mesh size is present as h^2 in our equation.

4)

```
Code: NE155 hw6 4
import matplotlib.pyplot as plt
import numpy as np
import Ne155 hw5 4 as It sol
import copy
SigA = 0.7
vSig = 0.6
h = 0.1
D = 1
a = 4
S = 8
def makeQf = (vSig, phi):
    Qf = vSig * phi
    return Qf
def makeA(h, n, D, SigA):
    row 1 = -D / (h ** 2)
    row 2 = (2 * D)/(h** 2) + SigA
    row 3 = -D/(h ** 2)
    values = [row_1, row_2, row_3]
    diag = [-1, 0, 1]
    A = sci.sparse.diags(values, diag, shape = (n+1, n+1)).todense()
    #Adjust A for boundary conditions
    A[0, 0] = 1
    A[0, 1] = 0
    A[n, n] = 1
    A[n, n-1] = 0
    return A
#phi in is the initial guess
def IterSOl(SigA, vSig, h, D, phi_in, tol, k_0, k_tol, absolute = False):
    n = np.size(phi_in)
    phi n = np.zeros(n)
    phi_old = phi_in / np.linalg.norm(phi_in)
    k new = 0
    k old = k 0
    A = makeA(h, n, D, SigA)
    Qf old = makeQf(vSig, phi in)
    phi_error = 0
    k = rror = 1
    max it = 100
    eigen_it = 0
    GS_it = [0]
```

```
while ((phi error > tol) and (eigen it < max it) and (k error > k tol)):
        eigen it = eigen it + 1
        (phi_n, GS_error, phi_it) = It_sol.Gauss_Seidel()
        #GS has been modified in hw 5 code to take in the inputs for this
problem
        GS it.append(phi it)
        Qf new = makeQf(vSig, phi n)
        k new = k old * (np.sum(Qf new) / np.sum(Qf old))
        if (absolute == True):
            phi error = np.linalq.norm(np.abs(phi n - phi old), 2)
            k = rror = np.abs(k new - k old)
        else:
            phi error = np.linalg.norm(np.abs(phi n - phi old), 2) /
np.linalg.norm(phi_new, 2)
            k = rror = np.abs(k new - k old)
        phi old = copy.deecopy(phi new)
        k old = copy.deecopy(k new)
        Qf old = copy.deecopy(Qf_new)
    phi n = phi n / np.linalq.norm(phi n)
return("k": k_new, "k_error": k_error, "phi": phi_n, "phi_error": phi_error, "GS Iterations": GS_it, "Power Iterations": eigen_it)
#Initialize guesses:
phi in = np.ones(n + 1)
k \ 0 = 1
tol = 1 * 10 **(-4)
k \text{ tol} = 1 * 10 **(-4)
sol = IterSol(SigA, vSig, h, D, phi in, tol, k 0, k tol, True)
print(sol["k"], sol["Power Iterations"])
x axis = np.linspace(-a, a, n + 1)
plt.plot(x axis, sol["phi"], 'r--x')
plt.xlabel('x')
plt.ylabel('phi(x)')
plt.title('Numerical Solution Eigenvector')
plt.show
Output:
0.812465570786
20
```

