

Homework assignment 1

Question 1:

Section a:

We will denote the discrete function as \mathbf{u} , and we use it as a vector of length $n = \left\lfloor \frac{\pi}{h} \right\rfloor$, such that $\mathbf{u}_i + O(h) = u(x_i)$ for $i = 0, \dots, n$. Now, from the Dirichlet BC we know that $\mathbf{u}_0 = 1$. Using central difference for second derivative we know that: $\frac{\mathbf{u}_{i+1} - 2\mathbf{u}_i + \mathbf{u}_{i-1}}{h^2} + O(h^2) = -4 \cos(2x)$.

Using Neumann BC with we can calculate:

$\frac{\mathbf{u}_{n+1} - \mathbf{u}_n}{h} = 0 \Rightarrow \mathbf{u}_n = \mathbf{u}_{n+1}$. From here we can define the Laplacian operator L'

$$\text{to be: } L' = \begin{pmatrix} -2 & 1 & 0 & \dots & \dots & 0 \\ 1 & -2 & 1 & 0 & \dots & 0 \\ 0 & 1 & \ddots & & & \\ \vdots & & \ddots & \ddots & & \\ \vdots & & & 1 & -2 & 1 \\ 0 & \dots & \dots & 0 & 1 & -2 \end{pmatrix}.$$

$$\text{We want to solve the linear equation: } L' \begin{pmatrix} \mathbf{u}_1 \\ \mathbf{u}_2 \\ \vdots \\ \mathbf{u}_{n-2} \\ \mathbf{u}_n \end{pmatrix} = \begin{pmatrix} -4 \cos(2x_1)h^2 - 1 \\ -4 \cos(2x_2)h^2 \\ \vdots \\ -4 \cos(2x_{n-2})h^2 \\ -4 \cos(2x_{n-1})h^2 - \mathbf{u}_n \end{pmatrix}.$$

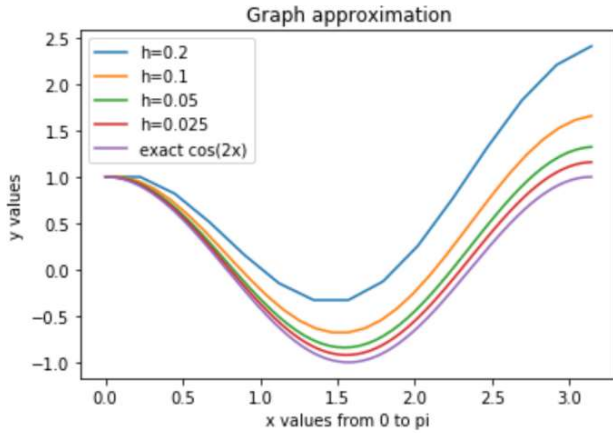
$$\text{So now we get: } L \begin{pmatrix} \mathbf{u}_1 \\ \mathbf{u}_2 \\ \vdots \\ \mathbf{u}_{n-1} \end{pmatrix} = \begin{pmatrix} -4 \cos(2x_1)h^2 - 1 \\ -4 \cos(2x_2)h^2 \\ \vdots \\ -4 \cos(2x_{n-1})h^2 \end{pmatrix}$$

$$\text{Where } L = \begin{pmatrix} -2 & 1 & 0 & \dots & \dots & 0 \\ 1 & -2 & 1 & 0 & \dots & 0 \\ 0 & 1 & \ddots & & & \\ \vdots & & \ddots & \ddots & & \\ \vdots & & & 1 & -2 & 1 \\ 0 & \dots & \dots & 0 & 1 & -1 \end{pmatrix}. \text{ (notice the difference in } L[n, n])$$

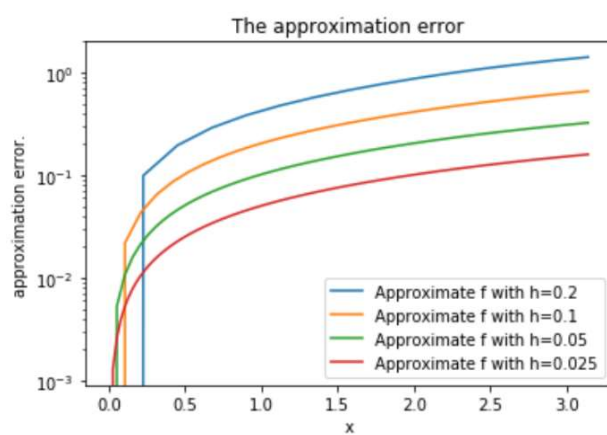
The code in Python will be as follows: (full code in zip file)

```
def numerical_sol_first_order(h):
    import numpy as np
    n = int(np.pi/h)
    actual_h = np.pi/(n-1)
    x = np.linspace(0, np.pi, n) #define the grid
    b = np.cos(2*x)*(-4)*(actual_h**2) #define solution vector
    n = x.size-1
    b[1] = 1 # Dirichlet BC
    main_diag = -2*np.ones(n)
    sub_diag = np.ones(n-1)
    L = np.diag(sub_diag, k=-1) + np.diag(main_diag) + np.diag(sub_diag, k=1)
    L[-1, -1] = 1 #Neumann BC
    s = np.linalg.solve(L, b[1:]) #solve the linear equation
    s = np.insert(s, 0, 1)
```

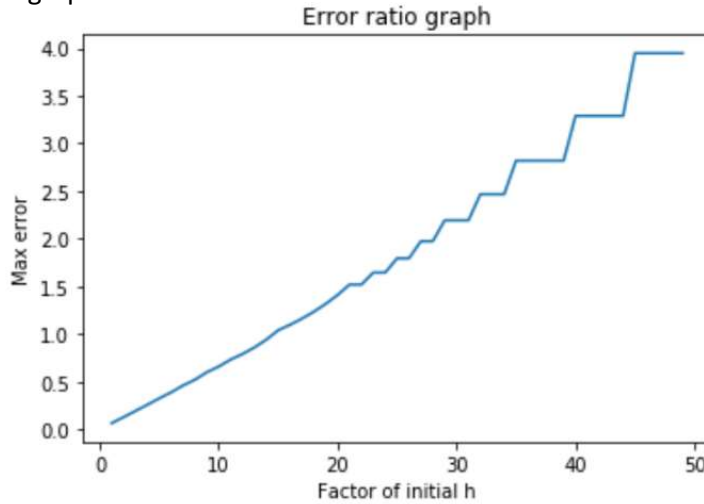
:Plotting the results we get:



Error graph in semi-logarithmic scale



Error ratio graph with initial $h=0.01$:



We can see the error grows in approximately linearity scale as h gets bigger, since we used first order approx.

Section b:

We will use second order central finite difference for Neumann BC.:

From the second derivative we know that: $-4 \cos(2x_n) = \frac{u_{n+1} - 2u_n + u_{n-1}}{h^2}$

and from the second order to Neumann BC we get $0 = \frac{u_{n+1} - u_{n-1}}{2h} \Rightarrow u_{n+1} = u_{n-1}$

Combining these equations we get: $-4 \cos(2x_n) h^2 = 2u_{n-1} - 2u_n$.

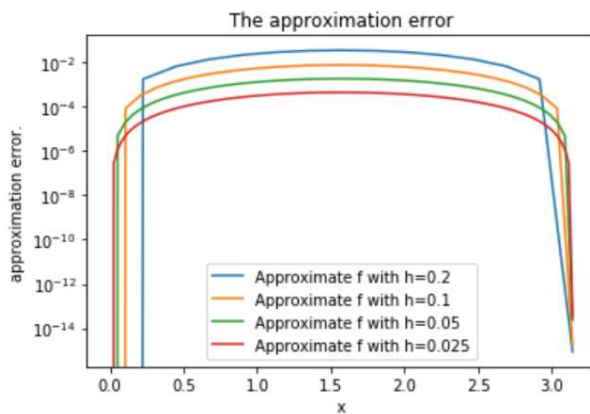
So now we can define our system of equations as follows:

$$\begin{pmatrix} -2 & 1 & 0 & \dots & \dots & 0 \\ 1 & -2 & 1 & 0 & \dots & 0 \\ 0 & 1 & \ddots & & & \\ \vdots & & \ddots & \ddots & & \\ \vdots & & & 1 & -2 & 1 \\ 0 & \dots & \dots & 0 & 2 & -2 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_{n-1} \\ u_n \end{pmatrix} = \begin{pmatrix} -4 \cos(2x_1) h^2 - 1 \\ -4 \cos(2x_2) h^2 \\ \vdots \\ -4 \cos(2x_{n-1}) h^2 \\ -4 \cos(2x_n) h^2 \end{pmatrix}$$

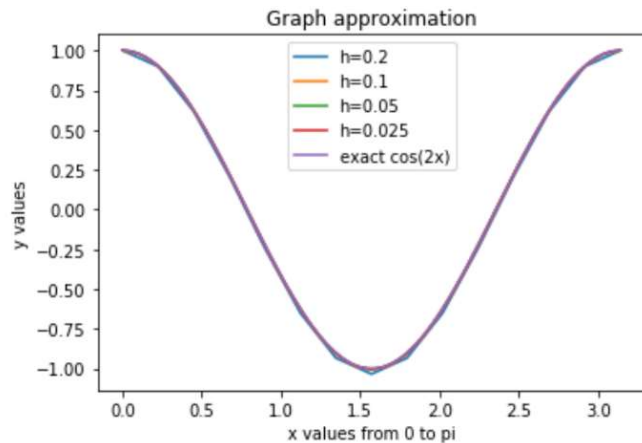
The code in Python will be as follows:

```
import numpy as np
def numerical_sol_second_order(h) :
    n = int(np.pi/h)
    actual_h = np.pi/(n-1)
    x = np.linspace(0,np.pi,n) #define the grid
    n=x.size-1
    b = np.cos(2*x)*(-4)*(actual_h**2) #define solution vector
    b[1]=-1 # Dirichlet BC
    main_diag = -2*np.ones(n)
    sub_diag = np.ones(n-1)
    L = np.diag(sub_diag,k=-1)+np.diag(main_diag) + np.diag(sub_diag, k=1)
    L[-1,-2]+=1 #Neumann BC
    s = np.linalg.solve(L,b[1:]) #solve the linear equation
    s = np.insert(s,0,1)
```

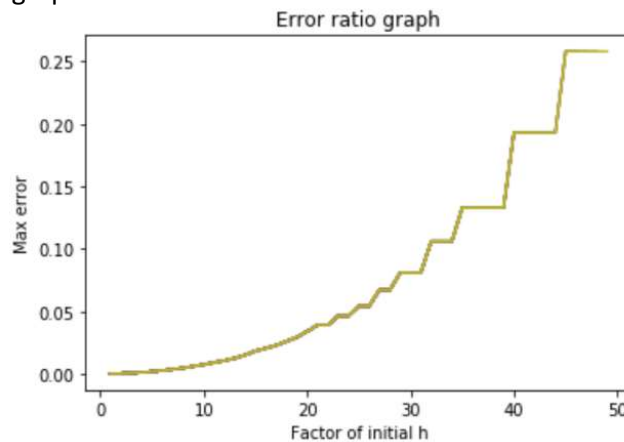
Error graph in semi logarithmic scale:



Plotting the results



Error ratio graph with initial h=0.01:



We can see that since we used second order approximation, the error grows in approximately square scale as h gets bigger.

Comparing the error graphs in semi logarithmic scale at both sections we can see that the in the second order approximation the solution is much more accurate. For instance, the max error in the first order for h=0.025 is bigger then the max order in the second order for only h=0.2.

Section c:

First, we will solve the ODE analytically:

By the laws of derivative we know: $\frac{\partial \left(\sigma(x) \left(\frac{\partial u}{\partial x} \right) \right)}{\partial x} = \sigma'(x) \frac{\partial u}{\partial x} + \sigma(x) \frac{\partial^2 u}{\partial^2 x} = q(x)$

so now we have: $\sigma'(x) \frac{\partial u}{\partial x} + \sigma(x) \frac{\partial^2 u}{\partial^2 x} = \sigma'(x)(-4 \cos(2x)) - \sigma(x)(2 \sin(2x))$

From here we can see that $\frac{\partial u}{\partial x} = -2 \sin(2x)$ and $\frac{\partial^2 u}{\partial^2 x} = -4 \cos(2x)$ satisfies

the solution. Now we get $\int \partial u = \int -2 \sin(2x) \partial x$ so $u = \cos(2x) + c$.

Putting together with the BC we get: $1 = \cos(0) + c \Rightarrow c = 0$.

So the solution to the ODE is $u(x) = \cos(2x)$.

Using the guide we calculate: $\frac{\partial u}{\partial x}(x_i) \approx \frac{u_{i+\frac{1}{2}} - u_{i-\frac{1}{2}}}{h}$, now we multiply the result with $\sigma(x)$ and apply the operator on the result:

$$\begin{aligned} \frac{\partial \left(\sigma(x) \left(\frac{\partial u}{\partial x} \right) \right)}{\partial x}(x_i) &\approx \frac{\sigma(x_{i+\frac{1}{2}})(u_{i+1} - u_i)}{h^2} - \frac{\sigma(x_{i-\frac{1}{2}})(u_i - u_{i-1}))}{h^2} = \\ &= \frac{1}{h^2} (\sigma(x_{i-\frac{1}{2}})u_{i-1} + (-\sigma(x_{i+\frac{1}{2}}) + \sigma(x_{i-\frac{1}{2}}))u_i + \sigma(x_{i+\frac{1}{2}})u_{i+1}). \end{aligned}$$

Using Dirichlet BC can approximate: $h^2 \cos(2x) = \sigma(x_{\frac{1}{2}})u_0 + (-\sigma(x_{\frac{3}{2}}) + \sigma(x_{\frac{1}{2}}))u_1 + \sigma(x_{\frac{3}{2}})u_2$

$$\Rightarrow (-\sigma(x_{\frac{3}{2}}) + \sigma(x_{\frac{1}{2}}))u_1 + \sigma(x_{\frac{3}{2}})u_2 = h^2 q(x_1) - \sigma(x_{\frac{1}{2}})$$

And similarly: $\sigma(x_{n-\frac{3}{2}})u_{n-2} + (-\sigma(x_{n-\frac{1}{2}}) + \sigma(x_{n-\frac{3}{2}}))u_{n-1} = h^2 q(x_{n-1}) - \sigma(x_{n-\frac{1}{2}})$

So we want to solve the following linear equation:

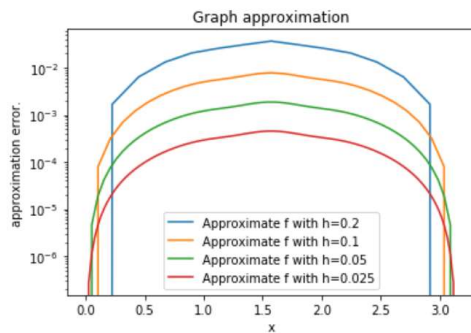
$$L \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_{n-2} \\ u_{n-1} \end{pmatrix} = \begin{pmatrix} q(x_1) - \sigma(x_{\frac{1}{2}}) \\ q(x_2)h^2 \\ \vdots \\ q(x_{n-2})h^2 \\ q(x_{n-1})h^2 - \sigma(x_{n-\frac{1}{2}}) \end{pmatrix} \text{ For } [L]_{i,j} = \begin{cases} -\sigma(x_{i+\frac{1}{2}}) + \sigma(x_{i-\frac{1}{2}}) & \text{if } i = j \\ \sigma(x_{i-\frac{1}{2}}) & \text{if } i = j + 1 \\ \sigma(x_{i+\frac{1}{2}}) & \text{if } i = j - 1 \\ 0 & \text{else} \end{cases}$$

The code in python will be as follows:

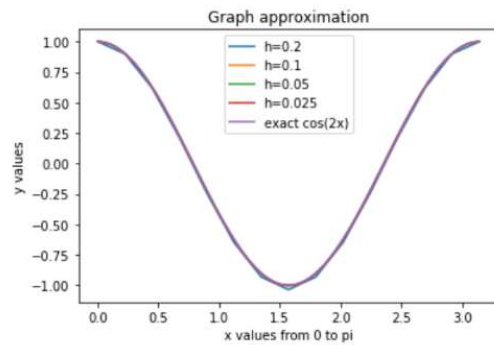
```
def sigma(x):
    return 1+(1/2)*np.exp(-10*((x-(0.5)*np.pi)**2))
def sigma_tag(x):
    return -10*(x-np.pi/2)*np.exp(-10*((x-(0.5)*np.pi)**2))
def f_q(x):
    return -4*sigma(x)*np.cos(2*x)-sigma_tag(x)*2*np.sin(2*x)

def numerical_sol_1c(h):
    n = int(np.pi/h)
    actual_h = np.pi/(n-1)
    x = np.linspace(0,np.pi,n) #define the grid
    b = (actual_h**2)*(f_q(x[1:-1]))#define solution vector
    b[0] = sigma(x[1]-actual_h/2) # Dirichlet BC
    b[-1] = sigma(x[-1]-actual_h/2)# Dirichlet BC
    main_diag = np.array([(-1)*(sigma(x[i]-0.5*actual_h)+sigma(x[i]+0.5*actual_h))for i in range(1,n-1)])
    sub_diag = np.array([sigma(x[i]+0.5*actual_h) for i in range(1,n-2)])
    L=np.diag(sub_diag,k=-1)+np.diag(main_diag) + np.diag(sub_diag, k=1)
    s = np.linalg.solve(L,b) #solve the linear equation
    s = np.concatenate((np.array([1]),s,np.array([1])))
```

Error graph in semi logarithmic scale:



Plotting the results



Error ratio with initial h=0.01:

