

APMA 922: Homework Set 02

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Problem 1

Part A

We begin by discretizing the equation of motion. Let $\Theta_i \equiv \theta(i\Delta t)$ where $\Delta t = T/n+1$ and n is the number of internal time points that we wish to solve for

$$\ddot{\theta}(t) + \sin(\theta(t)) = \frac{1}{h^2} [\Theta_{i+1} - 2\Theta_i + \Theta_{i-1}] + \sin(\Theta_i). \quad (1)$$

We define

$$G_i(\theta) \equiv \frac{1}{h^2} [\Theta_{i+1} - 2\Theta_i + \Theta_{i-1}] + \sin(\Theta_i) \quad (2)$$

as well as the associated Jacobian

$$J_{ij}(\theta) = \frac{1}{h^2} \begin{cases} 1, & j = i + 1 \text{ or } j = i - 1 \\ -2 + h^2 \cos \Theta_i, & j = i \\ 0, & \text{else.} \end{cases} \quad (3)$$

Then the k th iteration is given by

$$\theta^{k+1} = \theta^k + \delta \quad (4)$$

where δ is the solution to the linear system

$$J(\theta)\delta = -G(\theta^k). \quad (5)$$

This is implemented in the ipython notebook titled `A2Q1.ipynb`. The method is set up so that the algorithm continues the Newton refinement until either $\|\delta\| < \varepsilon_{32}$ or when the number of iterations is greater than 500, after which I assume that no solution would be found. Here, ε_{32} denotes the machine epsilon for single-precision. As can be seen in Fig. 1, this implementation reproduces successfully Figs. 2.4 and 2.5 in [LV].

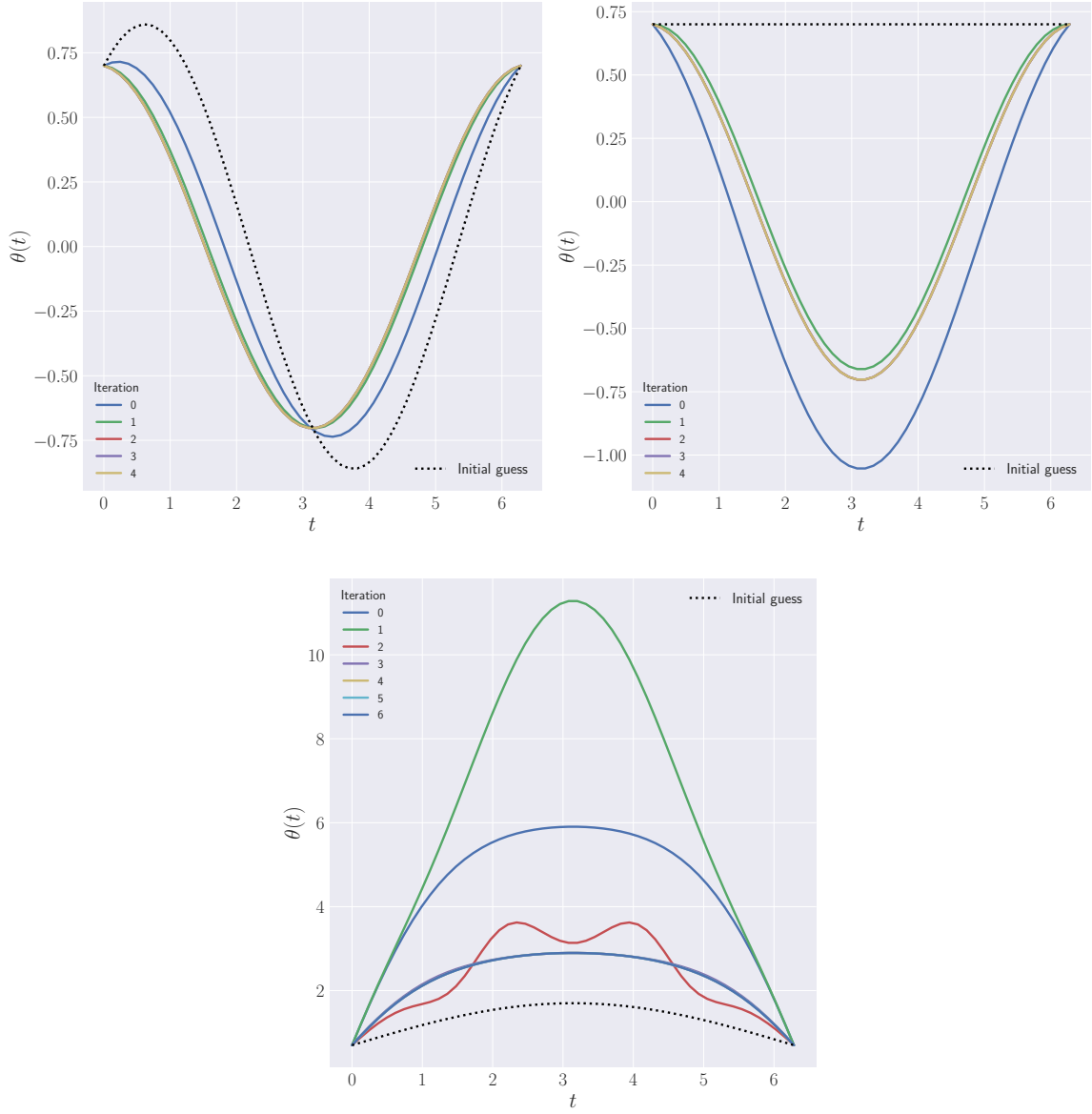


Figure 1: Recreation of Fig. 2.4 and Fig. 2.5 from [LV]. Angle θ as a function of time t . Initial guess is $\Theta_i^{[0]} = 0.7 + \sin(t_i/2)$. Different colors denote the iteration. Iteration counter starts from 0, not including initial guess. Initial guess is denoted by the black dotted line.

For an alternate initial solution $\theta^{[0]} = 0.7 \cos(t_i) + 2 \sin(t_i)$, I find an alternate solution shown in Fig. 2. This oscillation is similar to the one found in the top left-hand subplot of Fig. 1; however, this has a larger amplitude.

Part B

For longer time intervals, say $T = 20$, I find the following solution shown in Fig. ?? using the initial guess of $\theta^{[0]} = 0.9 \cos(t_i) + 0.2 \sin(t_i)$.

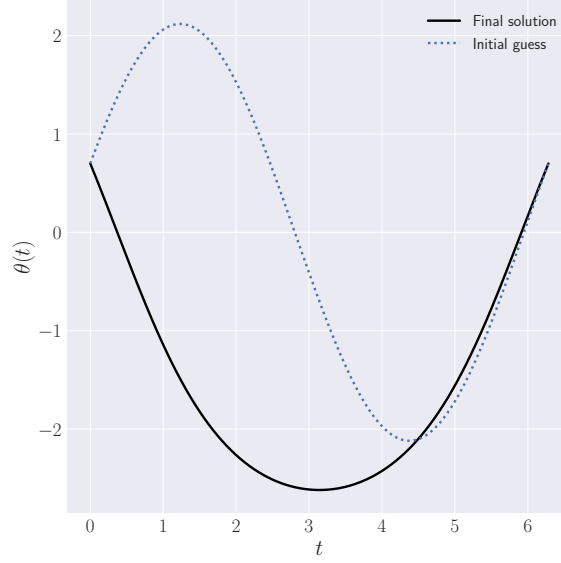


Figure 2: Solution of the Newton Iteration with initial guess $\theta^{[0]} = 0.7 \cos(t_i) + 2 \sin(t_i)$. Dashed blue line denotes the form of the initial guess, while the black solid line denotes the final solution found by the Newton iterator.

In general, I have not been able to find a generic solution which can have the Newton iteration able to find a solution. For a general guess that is not finely-tuned, the $\max \theta$ grows unbounded and behaves erratically, as can be seen in Fig. 4. Physically this does not make sense as a pendulum has multiple bounded orbits that intersect at $\theta = 0.7$ and thus the solutions that the Newton iterator finds which effectively explode are not valid. Intuitively, the asymptote of $\max \theta$ is π . I was also unable to find boundary layers, whose existence was suggested by this question.

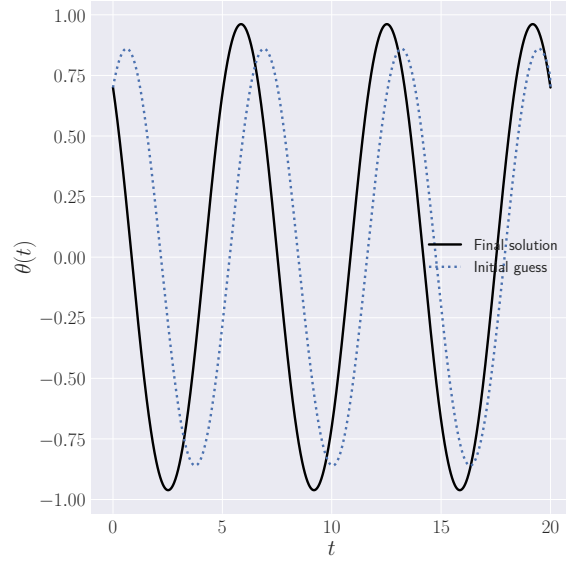


Figure 3: Solution found by the Newton Iteration for $T = 20$ using the initial solution $\theta^{[0]} = 0.9 \cos(t_i) + 0.2 \sin(t_i)$. Black solid line denotes the final solution after the convergence criterion was met. The dotted blue line denotes the initial guess input into the solver.

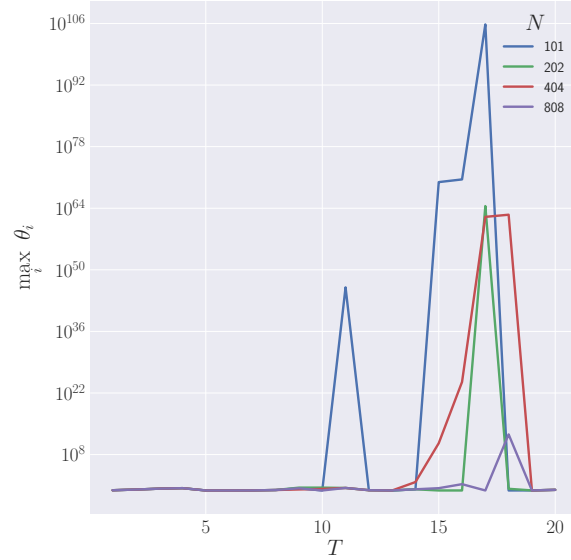


Figure 4: Solution found by the Newton Iteration for a range of T and for a variety of meshes with differing number of points m . Using the initial solution $\theta^{[0]} = 0.9 \cos(t_i) + 0.2 \sin(t_i)$. Different colors of lines denote different number of mesh points.

Problem 2

Part A

I implemented the five-point Laplacian by taking the code that was available on Canvas, translating it into Python, and then modifying the boundary conditions. This code can be found in the ipython notebook `A2Q2.ipynb`. Figure 5 shows the results of this implementation by reproducing all of the example figures in the provided MATLAB code.

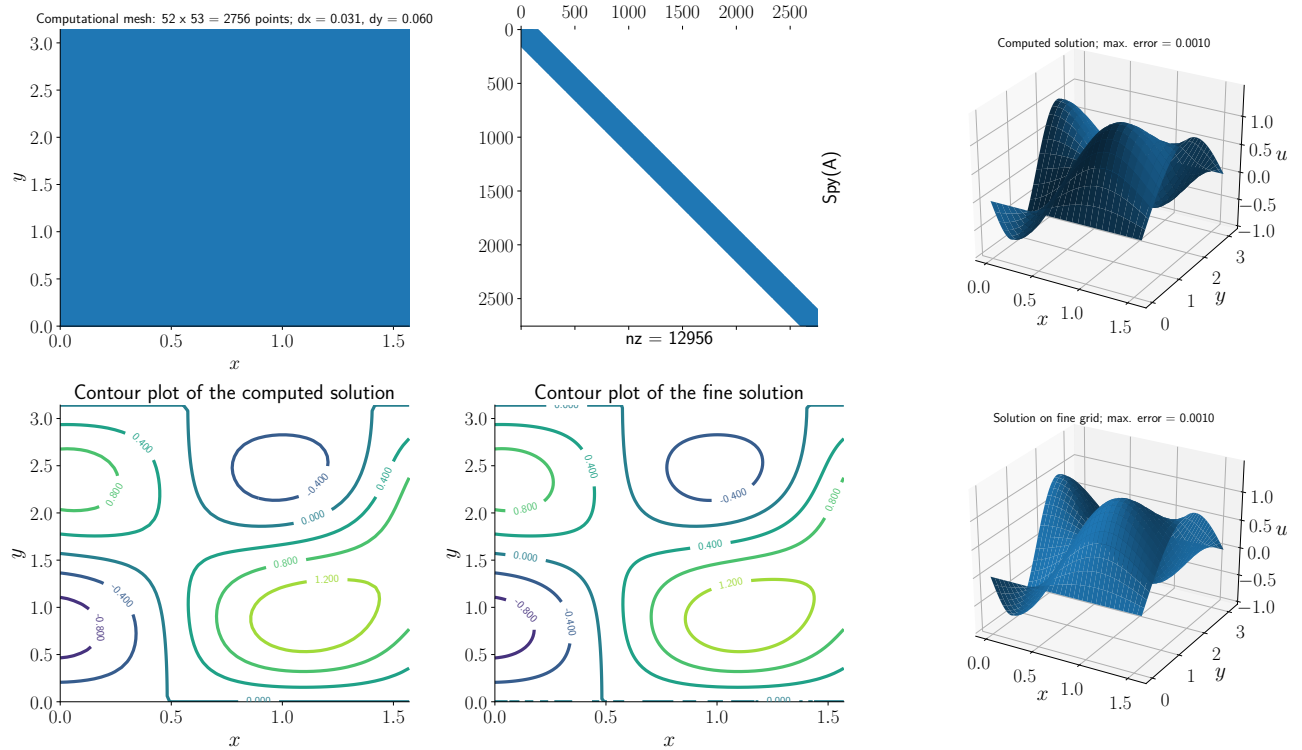


Figure 5: Recreation of the example plots for the differential equation $\nabla^2 u = 13 \cos 3x \sin 2y$, using the five-point Laplacian.

Performing grid refinement using this implementation of the five-point Laplacian, we observe in Fig. 6 that this implementation has accuracy of order two.

Part B

The nine-point Laplacian is given by the formula

$$\nabla_9^2 u_{ij} = \frac{1}{6h^2} [4(u_{i-1,j} + u_{i+1,j} + u_{i,j-1} + u_{i,j+1}) + (u_{i+1,j-1} + u_{i+1,j+1} + u_{i-1,j-1} + u_{i-1,j+1}) - 20u_{ij}] \quad (6)$$

To derive the truncation error we Taylor expand each of the u_{ij} . To simplify notation we

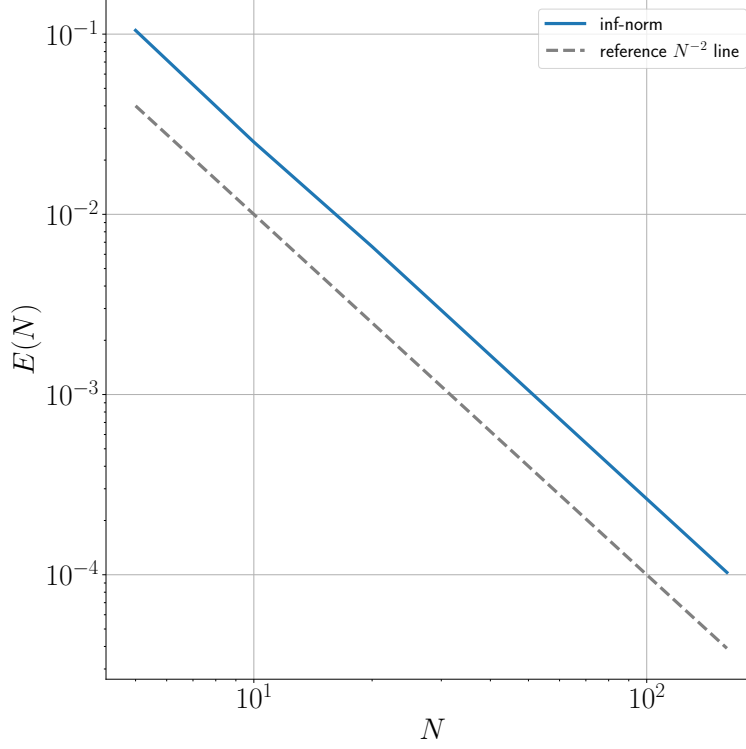


Figure 6: Error, $E(N)$, as a function of the number of mesh intervals, N , for the five-point Laplacian. We observe the desired order of two for this method.

define the shorthand,

$$u^{(m,n)}(\bar{x}, \bar{y}) \equiv \left. \frac{\partial^{m+n}}{\partial x^m \partial y^n} u(x, y) \right|_{x=\bar{x}, y=\bar{y}}. \quad (7)$$

Then, expanding each of the terms in (6), we have

$$u_{i-1,j} = u(\bar{x}, \bar{y}) - hu^{(1,0)}(\bar{x}, \bar{y}) + \frac{1}{2}h^2u^{(2,0)}(\bar{x}, \bar{y}) - \frac{1}{6}h^3u^{(3,0)}(\bar{x}, \bar{y}) + \frac{1}{24}h^4u^{(4,0)}(\bar{x}, \bar{y}) - \frac{1}{120}h^5u^{(5,0)}(\bar{x}, \bar{y}) + \mathcal{O}(h^6) \quad (8a)$$

$$u_{i+1,j} = u(\bar{x}, \bar{y}) + hu^{(1,0)}(\bar{x}, \bar{y}) + \frac{1}{2}h^2u^{(2,0)}(\bar{x}, \bar{y}) + \frac{1}{6}h^3u^{(3,0)}(\bar{x}, \bar{y}) + \frac{1}{24}h^4u^{(4,0)}(\bar{x}, \bar{y}) + \frac{1}{120}h^5u^{(5,0)}(\bar{x}, \bar{y}) + \mathcal{O}(h^6) \quad (8b)$$

$$u_{i,j-1} = u(\bar{x}, \bar{y}) - hu^{(0,1)}(\bar{x}, \bar{y}) + \frac{1}{2}h^2u^{(0,2)}(\bar{x}, \bar{y}) - \frac{1}{6}h^3u^{(0,3)}(\bar{x}, \bar{y}) + \frac{1}{24}h^4u^{(0,4)}(\bar{x}, \bar{y}) - \frac{1}{120}h^5u^{(0,5)}(\bar{x}, \bar{y}) + \mathcal{O}(h^6) \quad (8c)$$

$$u_{i,j+1} = u(\bar{x}, \bar{y}) + hu^{(0,1)}(\bar{x}, \bar{y}) + \frac{1}{2}h^2u^{(0,2)}(\bar{x}, \bar{y}) + \frac{1}{6}h^3u^{(0,3)}(\bar{x}, \bar{y}) + \frac{1}{24}h^4u^{(0,4)}(\bar{x}, \bar{y}) + \frac{1}{120}h^5u^{(0,5)}(\bar{x}, \bar{y}) + \mathcal{O}(h^6) \quad (8d)$$

Adding these equations to compute the first group of terms inside the bracket of (6) we have that

$$4(u_{i-1,j} + u_{i+1,j} + u_{i,j-1} + u_{i,j+1}) = 16u(\bar{x}, \bar{y}) + 4h^2 (u^{(0,2)}(\bar{x}, \bar{y}) + u^{(2,0)}(\bar{x}, \bar{y})) + \frac{1}{3}h^4 (u^{(0,4)}(\bar{x}, \bar{y}) + u^{(4,0)}(\bar{x}, \bar{y})) + \mathcal{O}(h^6) \quad (9)$$

Computing the expansions of the remaining terms we have

$$\begin{aligned} u_{i-1,j-1} = & u(\bar{x}, \bar{y}) + h(-u^{(0,1)}(\bar{x}, \bar{y}) - u^{(1,0)}(\bar{x}, \bar{y})) + \frac{1}{2}h^2 (u^{(0,2)}(\bar{x}, \bar{y}) + 2u^{(1,1)}(\bar{x}, \bar{y}) + u^{(2,0)}(\bar{x}, \bar{y})) \\ & - \frac{1}{6}h^3 (u^{(0,3)}(\bar{x}, \bar{y}) + 3(u^{(1,2)}(\bar{x}, \bar{y}) + u^{(2,1)}(\bar{x}, \bar{y})) + u^{(3,0)}(\bar{x}, \bar{y})) \\ & + \frac{1}{24}h^4 (u^{(0,4)}(\bar{x}, \bar{y}) + 4u^{(1,3)}(\bar{x}, \bar{y}) + 6u^{(2,2)}(\bar{x}, \bar{y}) + 4u^{(3,1)}(\bar{x}, \bar{y}) + u^{(4,0)}(\bar{x}, \bar{y})) \\ & - \frac{1}{120}h^5 (u^{(0,5)}(\bar{x}, \bar{y}) + 5(u^{(1,4)}(\bar{x}, \bar{y}) + 2(u^{(2,3)}(\bar{x}, \bar{y}) + u^{(3,2)}(\bar{x}, \bar{y})) + u^{(4,1)}(\bar{x}, \bar{y})) + u^{(5,0)}(\bar{x}, \bar{y})) \\ & + \mathcal{O}(h^6) \end{aligned} \quad (10a)$$

$$\begin{aligned} u_{i-1,j+1} = & u(\bar{x}, \bar{y}) + h(u^{(0,1)}(\bar{x}, \bar{y}) - u^{(1,0)}(\bar{x}, \bar{y})) + \frac{1}{2}h^2 (u^{(0,2)}(\bar{x}, \bar{y}) - 2u^{(1,1)}(\bar{x}, \bar{y}) + u^{(2,0)}(\bar{x}, \bar{y})) \\ & + \frac{1}{6}h^3 (u^{(0,3)}(\bar{x}, \bar{y}) - 3u^{(1,2)}(\bar{x}, \bar{y}) + 3u^{(2,1)}(\bar{x}, \bar{y}) - u^{(3,0)}(\bar{x}, \bar{y})) \\ & + \frac{1}{24}h^4 (u^{(0,4)}(\bar{x}, \bar{y}) - 4u^{(1,3)}(\bar{x}, \bar{y}) + 6u^{(2,2)}(\bar{x}, \bar{y}) - 4u^{(3,1)}(\bar{x}, \bar{y}) + u^{(4,0)}(\bar{x}, \bar{y})) \\ & + \frac{1}{120}h^5 (u^{(0,5)}(\bar{x}, \bar{y}) - 5u^{(1,4)}(\bar{x}, \bar{y}) + 5(2u^{(2,3)}(\bar{x}, \bar{y}) - 2u^{(3,2)}(\bar{x}, \bar{y}) + u^{(4,1)}(\bar{x}, \bar{y})) - u^{(5,0)}(\bar{x}, \bar{y})) \\ & + \mathcal{O}(h^6) \end{aligned} \quad (10b)$$

$$\begin{aligned} u_{i+1,j-1} = & u(\bar{x}, \bar{y}) + h(u^{(1,0)}(\bar{x}, \bar{y}) - u^{(0,1)}(\bar{x}, \bar{y})) + \frac{1}{2}h^2 (u^{(0,2)}(\bar{x}, \bar{y}) - 2u^{(1,1)}(\bar{x}, \bar{y}) + u^{(2,0)}(\bar{x}, \bar{y})) \\ & - \frac{1}{6}h^3 (u^{(0,3)}(\bar{x}, \bar{y}) - 3u^{(1,2)}(\bar{x}, \bar{y}) + 3u^{(2,1)}(\bar{x}, \bar{y}) - u^{(3,0)}(\bar{x}, \bar{y})) \\ & + \frac{1}{24}h^4 (u^{(0,4)}(\bar{x}, \bar{y}) - 4u^{(1,3)}(\bar{x}, \bar{y}) + 6u^{(2,2)}(\bar{x}, \bar{y}) - 4u^{(3,1)}(\bar{x}, \bar{y}) + u^{(4,0)}(\bar{x}, \bar{y})) \\ & - \frac{1}{120}h^5 (u^{(0,5)}(\bar{x}, \bar{y}) - 5u^{(1,4)}(\bar{x}, \bar{y}) + 5(2u^{(2,3)}(\bar{x}, \bar{y}) - 2u^{(3,2)}(\bar{x}, \bar{y}) + u^{(4,1)}(\bar{x}, \bar{y})) - u^{(5,0)}(\bar{x}, \bar{y})) \\ & + \mathcal{O}(h^6) \end{aligned} \quad (10c)$$

$$\begin{aligned} u_{i+1,j+1} = & u(\bar{x}, \bar{y}) + h(u^{(0,1)}(\bar{x}, \bar{y}) + u^{(1,0)}(\bar{x}, \bar{y})) + \frac{1}{2}h^2 (u^{(0,2)}(\bar{x}, \bar{y}) + 2u^{(1,1)}(\bar{x}, \bar{y}) + u^{(2,0)}(\bar{x}, \bar{y})) \\ & + \frac{1}{6}h^3 (u^{(0,3)}(\bar{x}, \bar{y}) + 3(u^{(1,2)}(\bar{x}, \bar{y}) + u^{(2,1)}(\bar{x}, \bar{y})) + u^{(3,0)}(\bar{x}, \bar{y})) \\ & + \frac{1}{24}h^4 (u^{(0,4)}(\bar{x}, \bar{y}) + 4u^{(1,3)}(\bar{x}, \bar{y}) + 6u^{(2,2)}(\bar{x}, \bar{y}) + 4u^{(3,1)}(\bar{x}, \bar{y}) + u^{(4,0)}(\bar{x}, \bar{y})) \\ & + \frac{1}{120}h^5 (u^{(0,5)}(\bar{x}, \bar{y}) + 5(u^{(1,4)}(\bar{x}, \bar{y}) + 2(u^{(2,3)}(\bar{x}, \bar{y}) + u^{(3,2)}(\bar{x}, \bar{y})) + u^{(4,1)}(\bar{x}, \bar{y})) + u^{(5,0)}(\bar{x}, \bar{y})) \\ & + \mathcal{O}(h^6). \end{aligned} \quad (10d)$$

Combining these terms to compute the remaining group of terms gives

$$\begin{aligned} u_{i+1,j-1} + u_{i+1,j+1} + u_{i-1,j-1} + u_{i-1,j+1} &= 4u(\bar{x}, \bar{y}) + 2h^2 (u^{(0,2)}(\bar{x}, \bar{y}) + u^{(2,0)}(\bar{x}, \bar{y})) \\ &\quad + \frac{1}{6}h^4 (u^{(0,4)}(\bar{x}, \bar{y}) + 6u^{(2,2)}(\bar{x}, \bar{y}) + u^{(4,0)}(\bar{x}, \bar{y})) + \mathcal{O}(h^6). \end{aligned} \quad (11)$$

Substituting (9) and (11) into (6) and simplifying yields

$$\nabla_9^2 u(\bar{x}, \bar{y}) = (u^{(0,2)}(\bar{x}, \bar{y}) + u^{(2,0)}(\bar{x}, \bar{y})) + \frac{1}{12}h^2 (u^{(0,4)}(\bar{x}, \bar{y}) + 2u^{(2,2)}(\bar{x}, \bar{y}) + u^{(4,0)}(\bar{x}, \bar{y})) + \mathcal{O}(h^4) \quad (12a)$$

$$= \nabla^2 u(\bar{x}, \bar{y}) + \frac{1}{12}h^2 \nabla^2 (\nabla^2 u(\bar{x}, \bar{y})) + \mathcal{O}(h^4) \quad (12b)$$

$$= \nabla^2 u(\bar{x}, \bar{y}) + \frac{1}{12}h^2 \nabla^4 u(\bar{x}, \bar{y}) + \mathcal{O}(h^4) \quad (12c)$$

Computing the truncation error then gives

$$\nabla_9^2 u(\bar{x}, \bar{y}) - \nabla^2 u(\bar{x}, \bar{y}) = \frac{1}{12}h^2 \nabla^4 u(\bar{x}, \bar{y}) + \mathcal{O}(h^4). \quad (13)$$

The nine-point Laplacian is only a valid approximation if the mesh width in both dimensions is equivalent. As such here we set the number of point m to specify the number of internal mesh points in the x-direction and from this we compute the necessary n number of internal mesh points in the y-direction such that the grid spacing in both dimensions are equivalent:

$$\Delta y = \Delta x \quad (14a)$$

$$\frac{L_y}{n+1} = \frac{L_x}{m+1} \quad (14b)$$

$$\implies n = \frac{L_y}{L_x} [m+1] - 1. \quad (14c)$$

Implementing this in the ipython notebook `A2Q2.ipynb`, in the section marked “Part B”. Figure 7, shows the recreation of the example plots using the nine-point Laplacian. We see that the implementation successfully reproduces the results acquired from the 5-point Laplacian.

Like the 5-point Laplacian, this implementation of the 9-point Laplacian yields an observed accuracy order of two, which can be seen in Fig. 8.

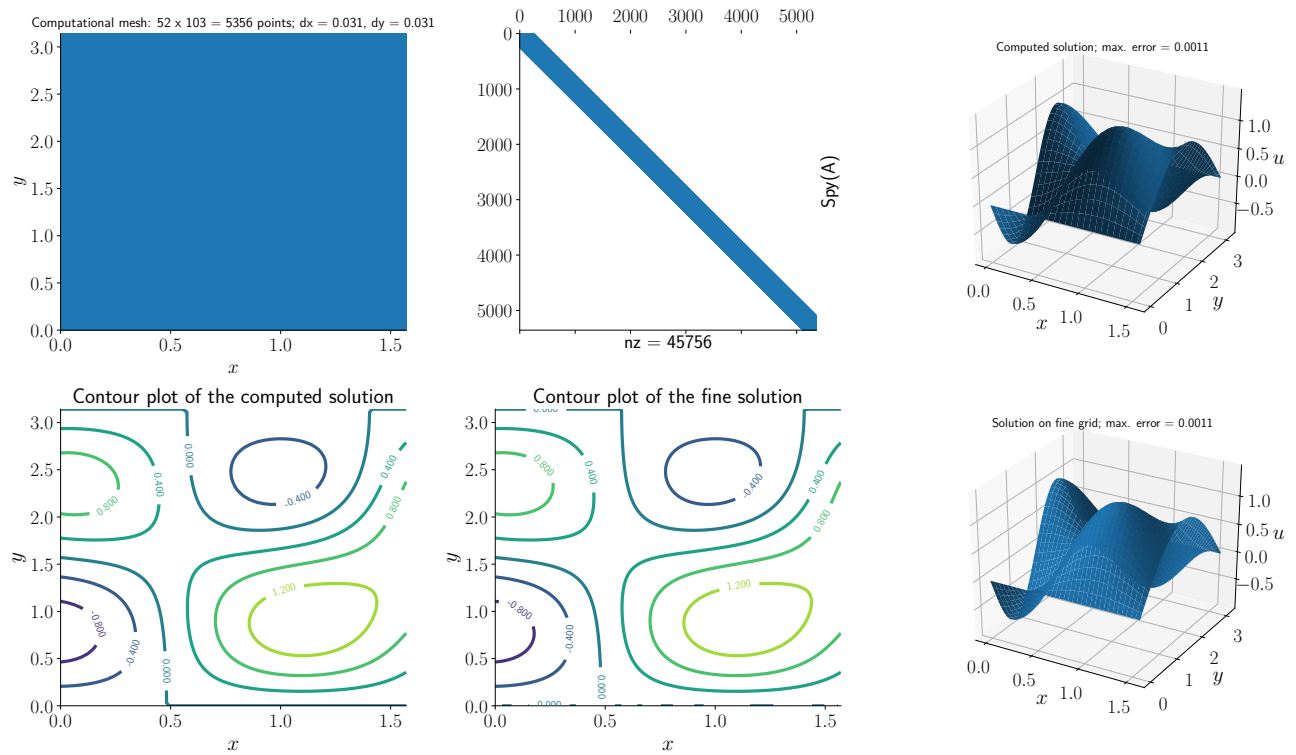


Figure 7: Recreation of the example plots using the 9-point Laplacian.

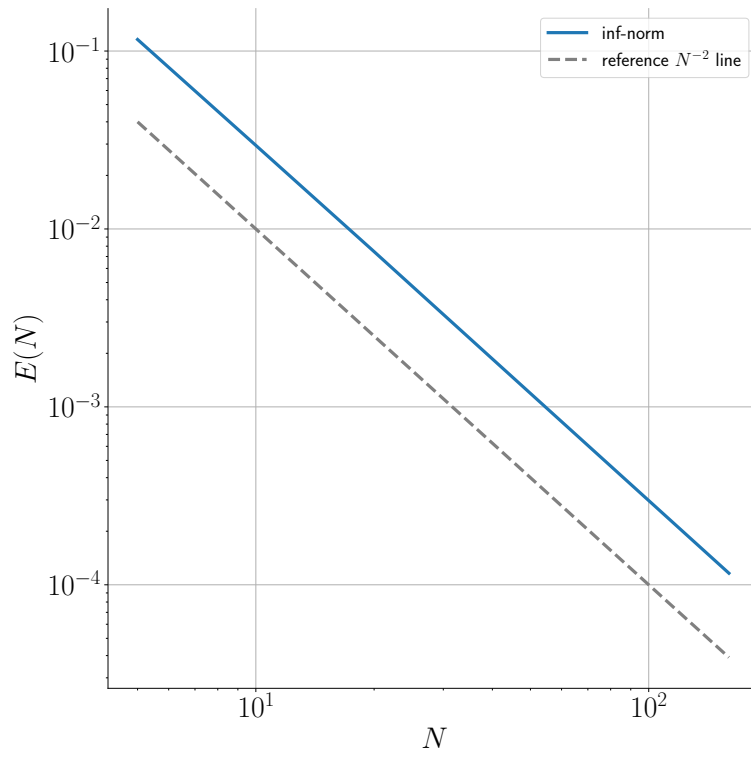


Figure 8: Error, $E(N)$, as a function of the number of mesh intervals, N , for the nine-point Laplacian. We observe an order of two for the accuracy for this method.

Implementing the deferred corrections by changing the right hand side of the PDE from,

$$f(x_i, y_j) \rightarrow \tilde{f}(x_i, y_j) \quad (15a)$$

$$= f(x_i, y_j) + \frac{h^2}{12} \nabla^2 f(x_i, y_j) \quad (15b)$$

$$= 13 \cos(3x) \sin(2y) - \frac{h^2}{12} 169 \cos(3x) \sin(2y), \quad (15c)$$

we acquire the error scaling observed in Fig. 9. We observe that this correction increases the order of the method by two from two to four.

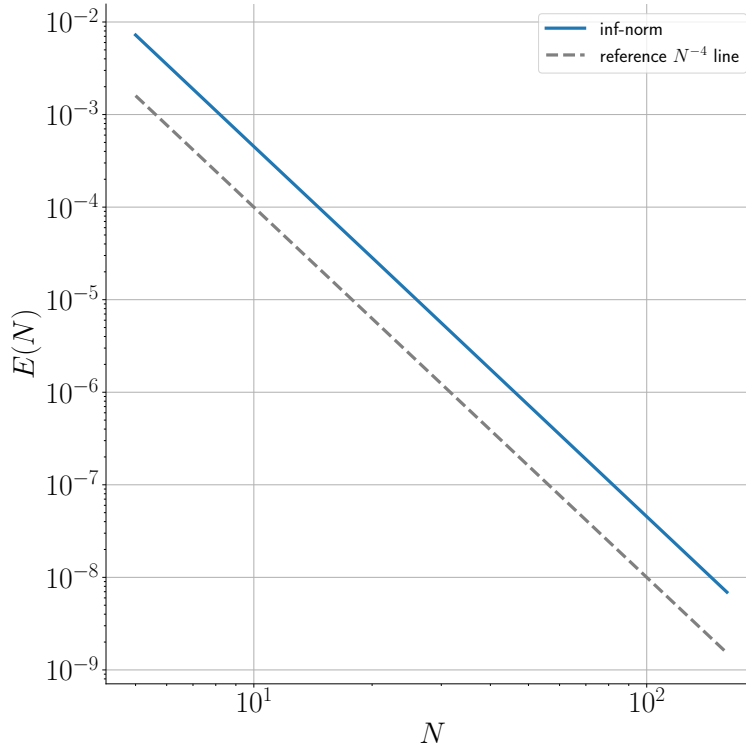


Figure 9: Error, $E(N)$, as a function of the number of mesh intervals, N , for the nine-point Laplacian with deferred corrections. We observe the desired order of four for this method.

Part C

We implement the Neumann Boundary condition using a second-order one-sided difference approximation so that

$$u_x(x = 0, y_j) \approx \frac{1}{h} \left[-\frac{3}{2} U_{0,j} + 2U_{1,j} - \frac{1}{2} U_{2,j} \right] = \frac{1}{2} \sin y_j. \quad (16)$$

Performing grid refinement on the implementation of the Neumann boundary condition using the five-point stencil, gives the desired order of two as can be seen in Fig. 10. I suspect that

the reason why this method deviates from the order 2 scaling at low N is because in those cases, the stencil of this boundary condition takes up the majority of points on the grid.

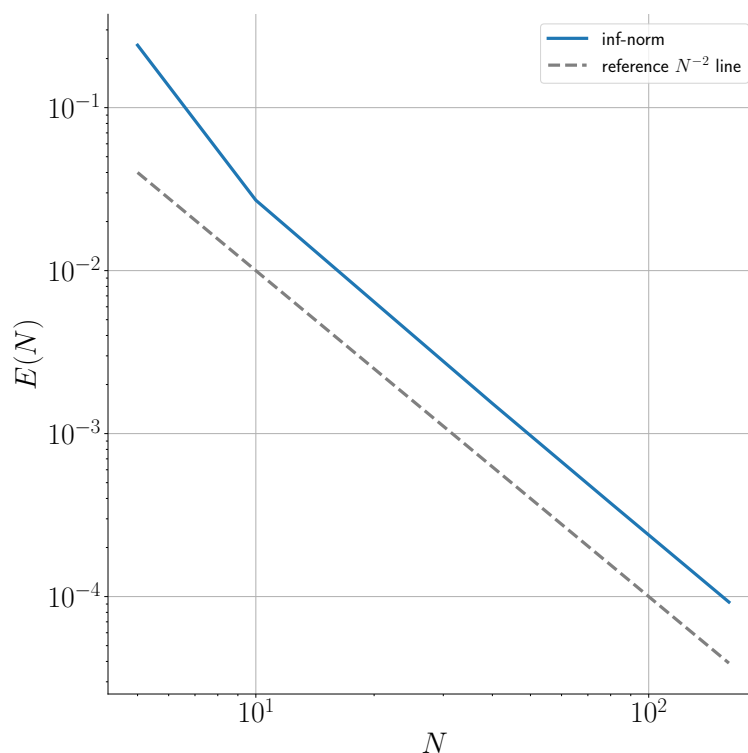


Figure 10: Error, $E(N)$, as a function of the number of mesh intervals, N , for the five-point Laplacian with Neumann boundary conditions at $x = 0$. We observe the desired order of two for this method.