Project 1 – Finite Differences and Solution Methods

Due: Oct. 14, 2020

Introduction to Problem Set 1

Problem set one is comprised of two questions and investigates finite difference methods for solving partial differential equations. The first question examines finite difference approximations on non-rectangular domains. The second question covers iterative solution methods, including Jacobi, Gauss Siedel, and multigrid techniques.

Problem 1 - Flow in a channel (55pts)

Problem Statement

We consider the problem of optimizing the cross section of a channel. The cross section, shown in Figure 1, is assumed to have a *constant* perimeter $l = 2(h - c) + 2\sqrt{b^2 + c^2}$, which is directly proportional to the amount of material required (for this problem l is assumed constant.) Therefore, the shape can be described in terms of two parameters/inputs, the height of the cross section h and the height of the angled portion of the base c.

Given c and h, the geometrical parameter b can be calculated from:

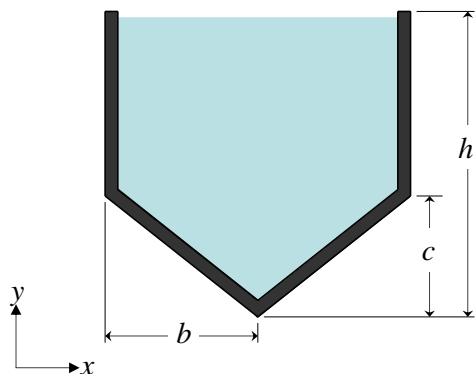$$b = \sqrt{\left(\frac{1}{2}l - h + c\right)^2 - c^2} \tag{1}$$
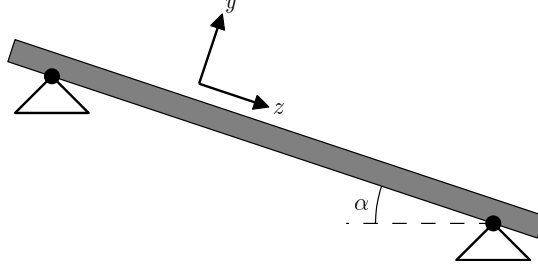

Figure 1: Channel cross section.

Figure 2: Flow channel.

We are interested in two outputs, the flow rate Q defined as the integral of the velocity u (normal to the cross section) over the cross section Φ

$$Q = \iint_{\Phi} u \, dx \, dy \tag{2}$$

and the moment of inertia of the cross section I. The last output, can be related to the deflection of the channel, due to the loading of the flowing fluid. The trapezoidal cross section has a thickness $t = 0.05$, and its moment of inertia I with respect to the neutral axis can be calculated from

$$I = 2\left(\frac{h+c}{2} - \bar{y}\right)^2 (h-c)t + \left(\bar{y} - c\bar{y} + \frac{1}{3}c^2\right) t\sqrt{b^2 + c^2}, \tag{3}$$

where the distance to the neutral axis is

$$\bar{y} = \frac{(h+c)(h-c) + c\sqrt{b^2 + c^2}}{2(h-c) + 2\sqrt{b^2 + c^2}}. \tag{4}$$

The channel is at a slope of angle α with respect to the horizontal direction, as shown in Figure 2, and the horizontal component of the gravitational force creates the pressure gradient for the downward flow. The governing equations are the steady Navier-Stokes equations,

$$\mathbf{u} \cdot \nabla \mathbf{u} + \nabla p = \mathbf{f} + \nu \nabla^2 \mathbf{u}. \tag{5}$$

With the assumption of fully developed flow, these can be reduced to Poisson's equation for the velocity component u normal to the channel cross section,

$$-\nabla^2 u = \frac{g \sin \alpha}{\nu}. \tag{6}$$

Here, g is the gravitational constant, and ν is the kinematic viscosity of the fluid. For simplicity, we assume that

$$\frac{g \sin \alpha}{\nu} = 1. \tag{7}$$

Along the walls of the channel the velocity is zero (no-slip condition), and on the free surface a zero-stress condition ($\frac{\partial u}{\partial n} = 0$) is assumed.

2

Numerical Procedure

We will solve Poisson's equation using a finite difference procedure. Given the symmetry of the geometry, we only consider half the channel section, as shown in Figure 3. The following boundary conditions are applied on the original domain:

$$\begin{cases} \frac{\partial u}{\partial n} = 0, & \text{in } AB \text{ and } AD \\ u = 0, & \text{in } BC \text{ and } CD. \end{cases} \tag{8}$$

To solve the problem, we map the half domain Ω to a rectangular domain $\hat{\Omega}$ and then solve the equations numerically using the finite difference method on $\hat{\Omega}$. We use a regular grid with $N \times N$ points in the computational domain, giving square cells of size $\Delta \xi = \Delta \eta = \frac{1}{N-1}$.
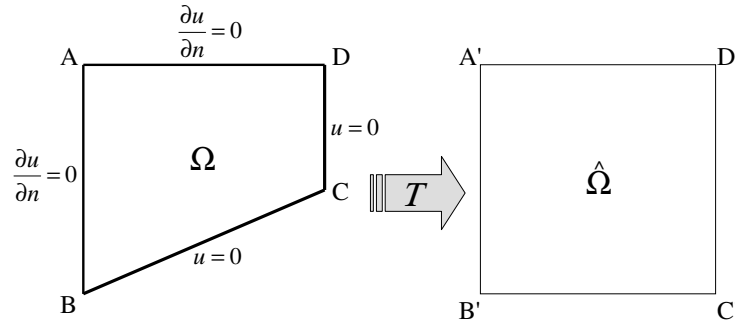

Figure 3: Geometrical transformation from the physical space to the computational domain.

Questions

1) (5pts) Introduce an analytical mapping T to transform the original domain Ω (with coordinates x, y) to the unit square computational domain $\hat{\Omega}$ (with coordinates ξ, η). Write the specific equation(s) you will solve and the corresponding boundary conditions in the computational domain.

2) (12pts) Use either Taylor series expansions, the method of undetermined coefficients or Lagrange's interpolant method to derive second-order accurate finite-difference schemes for the discretization of *all* the derivative terms in the interior of the domain, *as well as* for the boundary points. You do not have to derive special schemes for the corner points, use the top boundary scheme at corner A, and use $u = 0$ at the other three corners. Also show how you will numerically evaluate the integral for the approximate flow rate $\hat{Q}$.

3) (8pts) In a finite difference approximation, weighted combinations of the solution value at surrounding nodes are used to represent the governing differential equation at a given node. For node N, these weights are 'stamped' into appropriate positions in row N of the matrix that approximates the Laplacian operator in the domain. For a node $N_{i,j}$, present the nonzero entries of its corresponding matrix row. Include cases for interior **and** boundary nodes.
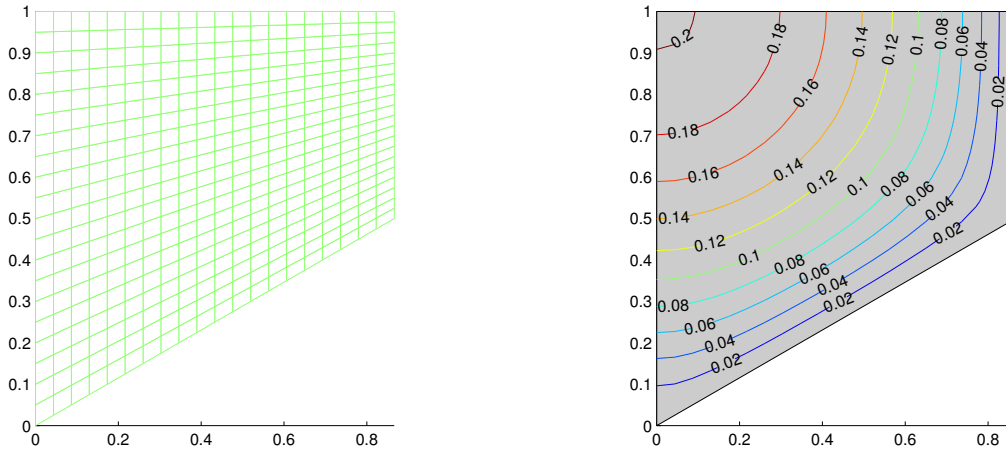
Figure 4: Sample solution, for $l = 3.0$, $c = 0.5$, $h = 1.0$, and $N = 21$. The plots show the grid (left) and contour lines of the solution (right). For these values, the flow rate $\hat{Q} = 0.129$.

4) (15pts) Write a program that solves the problem. A skeleton code for this problem has been provided in ChannelSkeleton.m. The use of the skeleton code is not mandatory. In addition one may decide to use any programming language to complete this assignment. Using the program you have developed, calculate the solution and $\hat{Q}$ using a grid size $N = 21$, for $l = 3.0$, $c = 0.5$, and $h = 1.0$. Compare your results with the sample solution in Figure 4. Please show your results and the comparison.

5) (3pts) Comment briefly(several bullets) on how the program you wrote to solve the channel flow problem would be extended to form a general Poisson solver for an arbitrary 4-sided solution domain.

6) (6pts) Calculate the convergence rate for the error in the output $\hat{Q}$, and for the L_∞ and the L_2 norms of the solution. Do the calculation for $l = 3.0$, $h = 1.0$, and $c = 0.2, 0.7$ (a total of 6 convergence plots); verify that you obtain second-order convergence. Use the grid sizes $N = 11, 21, 41, 81$. Since we don't know the exact solution, use the solution for $N = 81$ as a reference. To calculate the L_∞ and the L_2 norms, restrict the solution obtained at the reference mesh, to the current coarse mesh. Please plot your solutions on a semi-log or log-log plot where the errors are represented on the log scale. This is common practice, and should be applied to all future error plots even if it is not explicitly stated to do so.

7) (6pts) Keeping $l = 3$, vary the two inputs $h = [0.1, 1.0]$ and $c = [0.0, 1.0]$. Choose points on a regular grid in this two dimensional space, with constant interval 0.1 for both h and c. Each point, an (h, c) pair, represents a possible cross section. For each of these configurations, solve the problem (using $N = 21$) and calculate the flow rate $\hat{Q}$ and the moment of inertia I. Create a graph, using the two outputs and calculate the convex hull of these points. The convex hull, known as the Pareto optimal frontier, is a trade-off curve; for a specific choice of one output, we determine the optimal choice for the other output. Assuming we are interested in maximizing the flow rate, what can we say about the resulting geometry? Explain the results you obtain for the Pareto front.

Problem 2 - Iterative Methods: Jacobi, G-S, Multigrid (45pts (+10pts bonus))

Problem Statement

It is of interest to researching neuroscientists to determine which regions of the brain are active during a given neurological response. In order to non-invasively determine active parts of the brain, an electroencephalogram (EEG) is one tool which can be used. The EEG uses scalp mounted electrodes to measure the electrical pulses created by regions of the brain which are active. In addition to the electrode data, one can make assumptions pertaining to the value of the potential on the the scalp (the boundary of the domain). In order to determine which regions of the brain are active (sources of electrical activity), an inverse problem must be solved.

In addition to the EEG application, we could consider another example of an inverse problem. In this second example, a researcher may have an array of reaction chambers. Due to the constraints of the reactions being studied, it is possible that only the temperature and heat flux at the boundaries of the test chamber array can be measured. It is of interest to the research scientist to determine the positions in the array in which reactions have taken place. In order to determine the locations in the array which have undergone a reaction, the boundary measurements of the temperature and heat flux can be used, and an inverse problem may be solved to determine the heat source locations (which are equivalent to the array positions in which reactions have occurred).

In this question, we use a rudimentary representation and approach to 'solving' the inverse problem. Rather than directly solving the inverse problem (which in itself is an interesting problem), we consider the forward problem. In the forward problem, we assume the source positions and potential on the boundary are known, and calculate the corresponding boundary flux. We then compare the boundary flux for the given source arrangement with the known boundary flux for which we are trying to determine source positions. By trial and error, or by some more advanced methods and reasoning, we will eventually be able to determine the unknown source configurations by matching the known and computed boundary flux values.

To solve the equations, we will exercise the iterative solution techniques seen in class. The governing equation for the problem is Poisson's equation,

$$-\nabla^2 \phi(x, y) = f, \tag{9}$$

with boundary condition $\phi(x, y) = 0$, on the entire boundary. We will examine a unit square domain, see Figure 5.

The domain is divided into 36 block regions. The 36 blocks are arranged in a 6×6 array of blocks. The 16 blocks not on the domain boundary are labeled interior blocks. All blocks are set to have an $f = 0$, with the exception of four of the interior blocks which have $f = 1$. Those four interior blocks with $f = 1$ are said to be source blocks. The challenge in this particular problem is to determine the position of 4 source blocks given the distribution of the normal flux $\frac{\partial \phi}{\partial n} = g(x)$ on the boundary of the domain.

Questions

In order to solve the Poisson equation, we overlay a computational grid on the domain. For this problem, a 25×25 node computational grid is used unless otherwise specified. This would mean that a source block would correspond to a 5×5 set of nodes being set to $f = 1$.
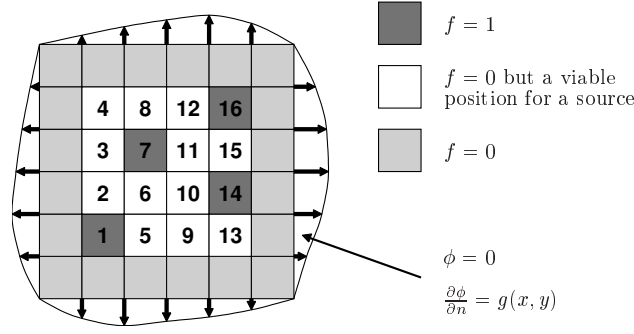
Figure 5: A pictorial explanation of the problem statement. Here the source blocks are located in blocks 1, 7, 14, and 16.

1) (8pts) Write down the Jacobi and the Gauss-Siedel iteration schemes to solve the above Poisson's equation.

2) (12pts) Code the Jacobi and G-S routines. Allow variable specification of the "fineness" of the computational grid (eg, a computational grid of 7×7 nodes, 13×13 nodes, 25×25 nodes, etc. while also allowing specification of the source blocks in the RHS vector).

   Note: For simplicity we shall assume that for the source blocks all nodes, including the block boundaries, have $f = 1$. Also, for two adjacent source blocks, the points on the common boundary also have $f = 1$.

   a. Implement a relaxation scheme for both. Describe the equation that you use, if you haven't already in the previous question.

   b. Show the convergence of your Jacobi and G-S solver for the above arrangement of blocks (1,7,14,16), with several relaxation factors. What is the optimal relaxation factor for the G-S method; this need not be a specific number, but a value within ± 0.1.

   c. A sample solution for the example case (1,7,14,16) is provided in Table 1 (left). Verify that your solver produces a similar output. Show either your matrix results or an image of your results.

   Hint: For later problems it will be convenient to construct a function that takes an input including the 4 numbers corresponding to the test blocks (B1, B2, B3, B4 $\in \{1,...,16\}$), and returns the normal derivative around the perimeter as an output.

3) (15pts) Implement a multigrid routine that will perform a two-grid method. Explain the restriction and prolongation method you choose to implement. Try a relaxation factor of $\frac{1}{2}$ and $\frac{4}{5}$. Please remember how you are expected to plot your error values from the previous problem (this will be expected to extend forward into all future assignments).

   a. Compare the convergence of the multigrid routine using either Gauss-Seidel, Jacobi, or both, with the various other methods. Which method is best? Which relaxation factor for your multigrid routine is better, $\frac{1}{2}$ or $\frac{4}{5}$?

   b. Implement routines where $\nu_1 = \nu_2 = 1$ and $\nu_c = 2, 4$, and exact (where ν_c refers to the number of iterations on the coarse grid), for the multigrid. How do the routines compare? What can we conclude from this?

4) (10pts) Using the best method (based on convergence rate) and the given normal flux distribution, determine the unknown positions of the four blocks, for the $\frac{\partial \phi}{\partial n} = g(x)$ given in Table 1 (right). You do not need to find an elaborate/efficient way to do it; you can just try all the possible combinations. Pictorially show where the sources are located.

5) BONUS: (Possible +10pts[1]) Write a generalized V-cycle multigrid routine which allows multiple grid refinements. Show convergence for the 2, 3, and 4 grid refinement V-cycles. How do they compare to the other iterative routines?

Numerical Values of Normal Flux Distribution

(1,7,14,16) Configuration

Pt	Left	Right	Bottom	Top
1	0.0000	0.0000	0.0000	0.0000
2	0.0122	0.0072	0.0122	0.0054
3	0.0244	0.0145	0.0244	0.0109
4	0.0361	0.0222	0.0360	0.0163
5	0.0466	0.0303	0.0463	0.0217
6	0.0552	0.0391	0.0548	0.0270
7	0.0612	0.0483	0.0605	0.0323
8	0.0643	0.0576	0.0633	0.0373
9	0.0649	0.0662	0.0635	0.0420
10	0.0636	0.0734	0.0618	0.0464
11	0.0611	0.0784	0.0590	0.0504
12	0.0583	0.0811	0.0561	0.0542
13	0.0553	0.0819	0.0533	0.0578
14	0.0522	0.0814	0.0507	0.0612
15	0.0489	0.0803	0.0483	0.0645
16	0.0452	0.0790	0.0459	0.0673
17	0.0411	0.0771	0.0431	0.0689
18	0.0367	0.0740	0.0399	0.0684
19	0.0319	0.0687	0.0360	0.0652
20	0.0268	0.0610	0.0315	0.0589
21	0.0216	0.0509	0.0262	0.0498
22	0.0162	0.0392	0.0202	0.0387
23	0.0108	0.0264	0.0138	0.0262
24	0.0054	0.0132	0.0070	0.0132
25	0.0000	0.0000	0.0000	0.0000

Unknown Configuration

Pt	Left	Right	Bottom	Top
1	0.0000	0.0000	0.0000	0.0000
2	0.0083	0.0059	0.0082	0.0050
3	0.0167	0.0117	0.0161	0.0100
4	0.0254	0.0175	0.0238	0.0148
5	0.0345	0.0233	0.0310	0.0195
6	0.0439	0.0289	0.0376	0.0240
7	0.0534	0.0343	0.0434	0.0282
8	0.0626	0.0394	0.0483	0.0322
9	0.0704	0.0441	0.0521	0.0360
10	0.0761	0.0483	0.0549	0.0397
11	0.0787	0.0520	0.0566	0.0432
12	0.0780	0.0553	0.0572	0.0468
13	0.0743	0.0582	0.0566	0.0504
14	0.0683	0.0610	0.0551	0.0542
15	0.0611	0.0636	0.0525	0.0579
16	0.0536	0.0657	0.0490	0.0612
17	0.0462	0.0667	0.0448	0.0633
18	0.0393	0.0658	0.0400	0.0635
19	0.0328	0.0624	0.0347	0.0608
20	0.0267	0.0561	0.0292	0.0552
21	0.0209	0.0473	0.0234	0.0468
22	0.0155	0.0365	0.0176	0.0363
23	0.0102	0.0247	0.0117	0.0246
24	0.0051	0.0124	0.0059	0.0124
25	0.0000	0.0000	0.0000	0.0000

Table 1: These are the tabulated results for the normal flux, for the (1,7,14,16) configuration (left) and for the unknown configuration (right), for which you are to solve. The points correspond to the grid points in a 25×25 node grid, for increasing y-values (left and right boundaries), and for increasing x-values (bottom and top boundaries).

---

[1]Only applied to gain a maximum of 100%. Additional bonus points are not carried over to future assignments.

MATLAB **Hints**

- It is usually necessary to *vectorize* MATLAB code in order to get high performance. This means, for example, that for-loops should be avoided and replaced by higher-level constructs. We *do not* require you to do this, feel free to use for-loops when building up matrices and vectors. This will make the code easier to read and understand, and these (relatively) small problems are fast enough anyway.

- A uniform grid with spacings `dxi`, `deta` can be created with

      [xi,eta]=ndgrid(0:dxi:1,0:deta:1);

  This gives two $N \times N$ arrays `xi`, `eta`.

- A grid with x, y coordinates in the $N \times N$ arrays `x`, `y` can be visualized using

      mesh(x,y,0*x,0*x)
      view(2), axis equal

- Remember to make A a *sparse matrix* (or you will run out of memory): `A=sparse(N^2,N^2);`

- When filling in the A matrix, it is convenient to use a mapping function, `map=reshape(1:N^2,N,N);` The position of the grid point `i, j` in the linear system of equations is then `map(i,j)`.

- The solution U to the linear system of equations $AU = F$ can be computed with `U=A\F`. If `A` is a sparse matrix, this will use the direct sparse solver in MATLAB.

- The $N^2 \times 1$ solution vector `U` can be reshaped to an $N \times N$ array with `u=reshape(U,N,N);` This format is sometimes easier to work with, for example when computing the integral for $\hat{Q}$ and when plotting the solution.

- The contour plot in Figure 4 was created with

      patch([0,b,b,0],[0,c,h,h],-ones(1,4),0,'facecolor',[.8,.8,.8])
      [cc,hh]=contour(x,y,u,0.02:0.02:max(u(:)));
      clabel(cc,hh);
      axis equal

  First, the geometry is drawn, using the parameters `b`, `c`, and `h`. Then contour curves are made for the solution `u` on the grid `x`, `y`, all which are $N \times N$ arrays.

- Alternatively, you can create a color plot:

      surf(x,y,0*x,u)
      view(2),axis equal
      set(gcf,'renderer','zbuffer');
      shading interp, colorbar

- To plot the convex hull for the points in the vectors `Qs` and `Is`, type

      k=convhull(Qs(:),Is(:));
      plot(Qs(:),Is(:),'.',Qs(k),Is(k))