FLORIDA INSTITUTE OF TECHNOLOGY MECHANICAL AND AEROSPACE ENGINEERING DEPARTMENT

MAE 5150-E1: Computational Fluid Dynamics

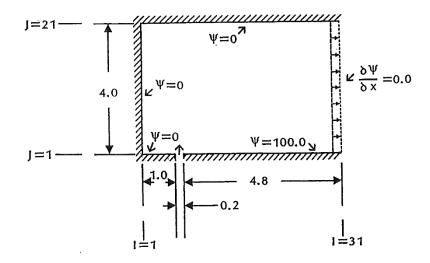
Fall 2017

Coding Project 2

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A two-dimensional inviscid, incompressible fluid is flowing steadily through a chamber between the inlet and the outlet, as shown in the figure. It is required to determine the streamline pattern within the chamber.



For a two-dimensional, incompressible flow, the continuity equation is expressed as

$$\frac{\partial u}{\partial x} + \frac{\partial v}{\partial y} = 0$$

A stream function Ψ may be defined such that

$$u = \frac{\partial \Psi}{\partial v}$$
 and $v = -\frac{\partial \Psi}{\partial x}$

Recall that a streamline is a line of constant stream function. Furthermore, vorticity is defined as

 $\Omega = \nabla \times \qquad V$

for which

$$\Omega_z = \frac{\partial v}{\partial x} - \frac{\partial u}{\partial y}$$

For an irrotational flow, the vorticity is zero. Therefore,

$$\frac{\partial v}{\partial x} - \frac{\partial u}{\partial y} = 0$$

Substituting the definitions of the stream function into the above equations yields

$$\frac{\partial}{\partial x} \left(-\frac{\partial \Psi}{\partial x} \right) - \frac{\partial}{\partial y} \left(\frac{\partial \Psi}{\partial y} \right) = 0$$

or

$$\frac{\partial^2 \Psi}{\partial x^2} + \frac{\partial^2 \Psi}{\partial y^2} = 0$$

The goal in this problem is to obtain the solution of this elliptic partial differential equation using the various numerical techniques discussed earlier. The solution will provide the streamline pattern within the chamber.

Since the chamber walls are streamlines, i.e. lines of constant Ψ , we will assign values for these streamlines, as shown in the figure. Solve this problem with codes using the following techniques:

- a) Point Gauss-Seidel
- c) Point SOR
- b) Line Gauss-Seidel
- d) Line SOR

For all methods, the step sizes are specified as

$$\Delta x = 0.2,$$
 $\Delta y = 0.2$ $ERRORMAX = 0.01$

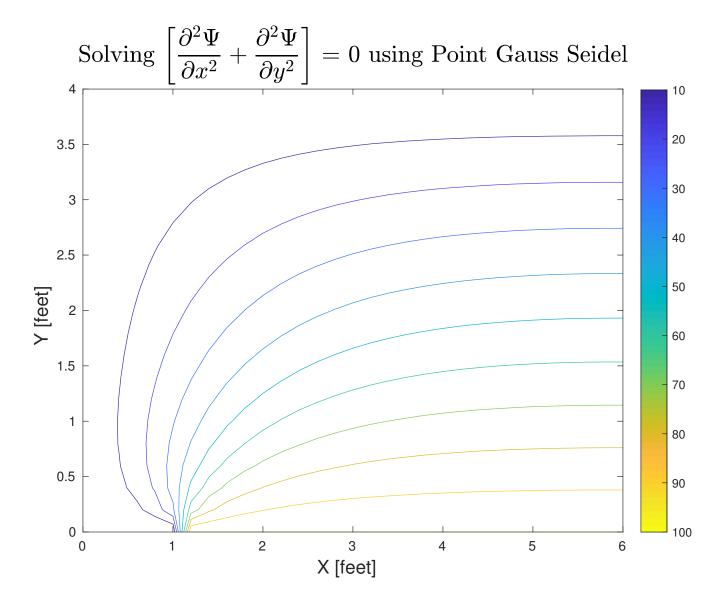
with convergence criterion *ERROR* < *ERRORMAX*, where

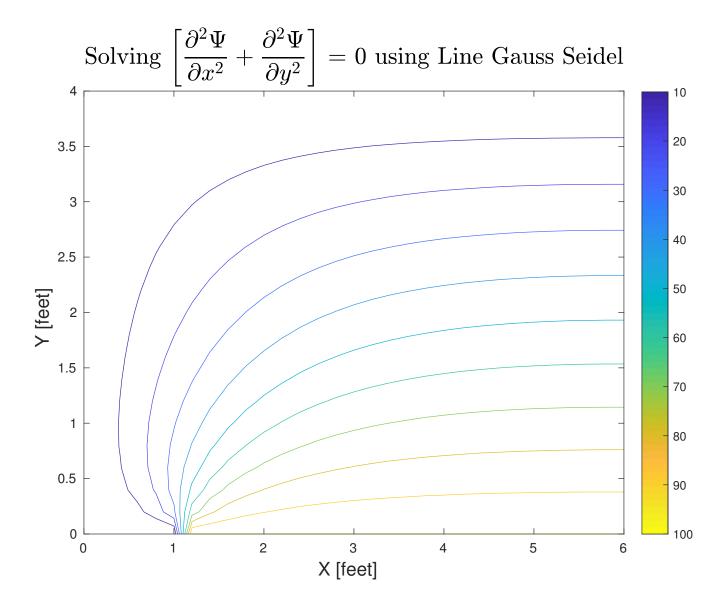
$$ERROR = \sum_{\substack{i=1 \\ i=2}}^{j=JM-1} \left| \Psi_{i,j}^{k+1} - \Psi_{i,j}^{k} \right|$$

Use an initial data distribution of $\Psi=0.0$. Plot the streamline pattern (lines of constant). Rerun the SOR codes for several values of the relaxation parameter and plot the relaxation parameter versus the number of iterations for these two schemes. In each case, determine the optimal value of the relaxation parameter.

Submit a both a hardcopy of your code and an electronic copy on Canvas.

Note: Except for the purpose of creating plots, MATLAB, Excel, or other commercially available software may not be used for this assignment. You may otherwise program in any language you wish.





W	Iteration	Time[sec]
W 0.100 0.200 0.300 0.400 0.500 0.600 0.700 0.800 0.900 1.000 1.100 1.200 1.300 1.400 1.500 1.600 1.700 1.800	Iteration 	Time[sec]
1.900	99 39686	0.002697 1.085977

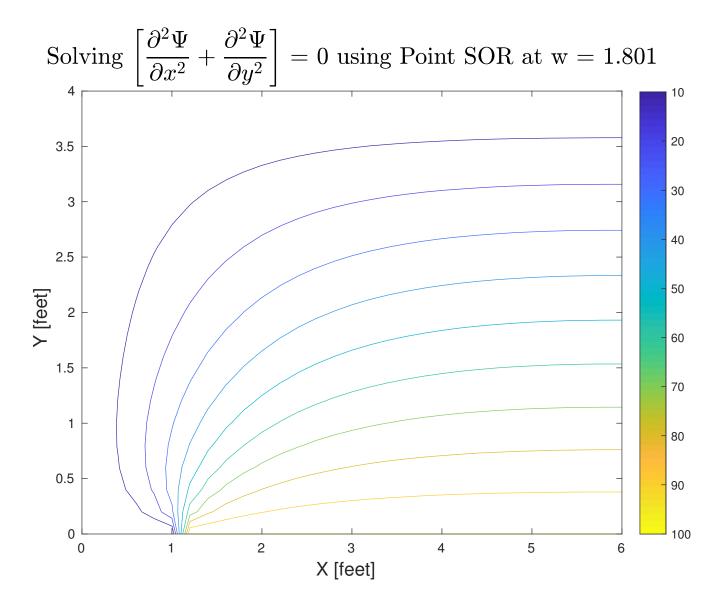
Table 1- Different values of relaxation parameter (w) for Point SOR

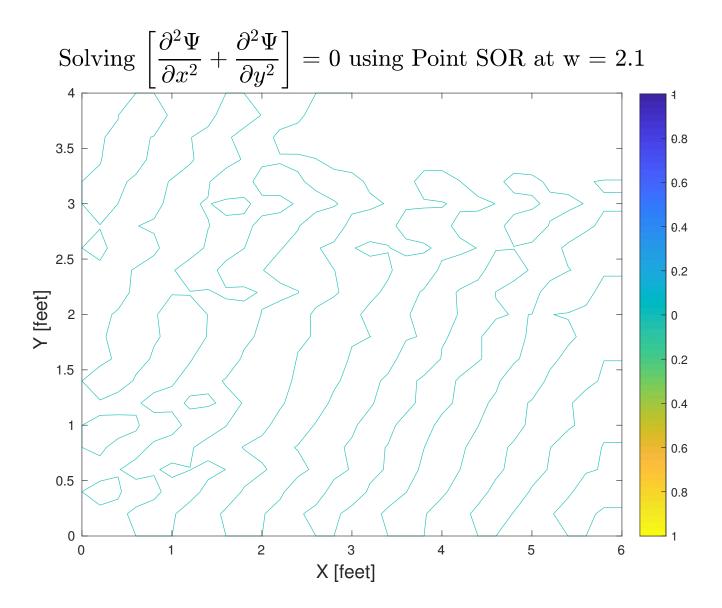
It can be seen that at around w = 1.8 and w = 1.9, the solution converges faster (less iterations and less computational time). After testing with different values between 1.8 and 1.9, it is decided that w = 1.801 is the optimum relaxation parameter. In the code, setting w = 1.801 takes 91 iterations and 0.002460 seconds.

W	Iteration	Time[sec]
W 0.100 0.200 0.300 0.400 0.500 0.600 0.700 0.800 0.900 1.000 1.100	Iteration 	Time[sec]
1.200 1.300 1.400 1.500 1.600 1.700 1.800 1.900 2.000	171 83 2564 566 334 232 170 121 61	0.006466 0.003198 0.095974 0.021152 0.012560 0.009338 0.007004 0.005359 0.002851

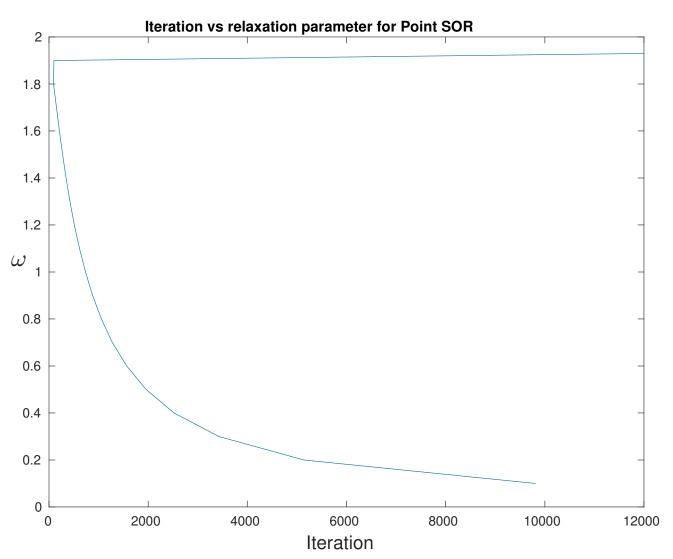
Table 2- Different values of relaxation parameter (w) for Line SOR

A similar observation can be said for the Line SOR case: at $\mathbf{w} = \mathbf{1.3}$, the solution converges the fastest (at 83 iterations and 0.003198 seconds). It can also be seen from later plots of relaxation parameter vs. iteration numbers, that anything higher than the optimum value will not guarantee convergence. As a result, $\mathbf{w} = 1.3$ is chosen as the optimal value for Line SOR.

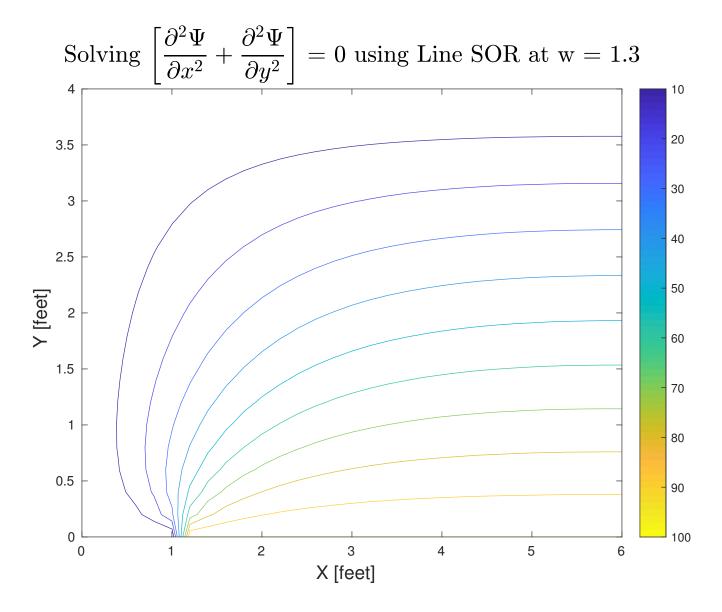


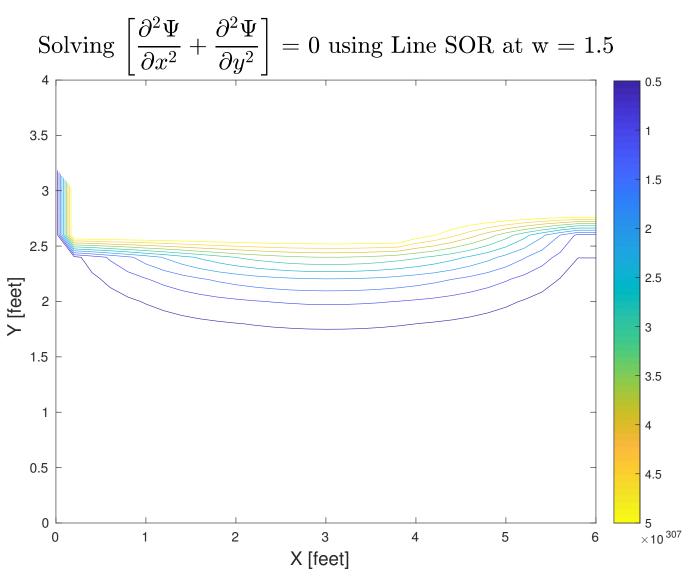


At w>2, the solution blows up and diverges instead.

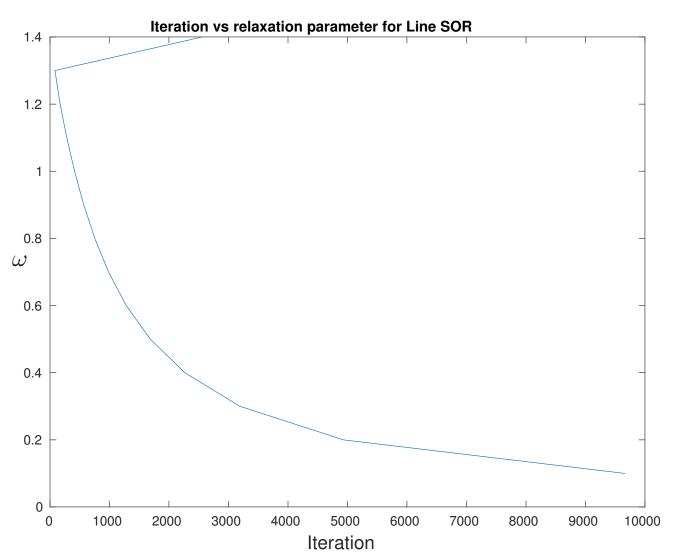


It can be seen that higher W will give lower Iterations. However, very high W will cause the Iterations to rise and from later plots, the solution can also diverge. The optimum relaxation parameter is shown clearly in this plot, where 1.8<w<1.9.





For Line SOR, the solution reaches optimum convergence at a lower w than Point SOR. Still, passing this optimal value causes divergence. We can see that the code tries to keep the derivative boundary condition (dPsi/dx = 0) on the right hand side, but due to the divergence of earlier data points, this does not help it to converge.



Line SOR converges faster than Point SOR, and also reaches optimal value of relaxation parameter faster. The same trend can be observed: higher W will give lower Iterations, but very high W will cause divergence. It can also be seen clearly that w = 1.3 is the optimum value for this scheme (less iterations).