



Programming Assignment #2 : Solving 1D Wave Equation

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November 15, 2016

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1. OVERVIEW

There are many physical phenomena in physics and engineering that require linear and non-linear partial differential equations to describe the true nature of the system. Solving these systems analytically and finding exact solutions for these equations can be difficult and often require simplifications that ultimately don't fully represent the problem being investigated. Numerical methods for solving PDEs provide a means of finding approximations to the exact solutions without having to make sacrificial simplifications. With recent advancements in computational technology, numerical methods can now be easily applied to large and difficult problems that would otherwise be impossible to solve.

1.1. THEORY

Numerical problems are essentially solved by breaking the entire solution domain into small discrete points (mesh) and finding the solution at or around these areas. Each point requires solving the differential equations that represent the physical phenomenon being investigated. Since the exact solution cannot be computed, it is instead approximated using various techniques and methods.

The Runge-Kutta scheme is a family of numerical techniques used to solve ordinary differential equations by numerically integrating using trial steps at the midpoints of an interval to cancel out lower-order error terms. A two-stage Runge-Kutta scheme is displayed in equation 1.2 which uses Explicit Euler as the first stage intermediate step 1.1.

$$w^{(1)} = w^n + \Delta t(\lambda w^n) \quad (1.1)$$

$$w^{n+1} = w^n + \Delta t \left(\frac{\lambda w^n + \lambda w^{(1)}}{2} \right) \quad (1.2)$$

The program will implement a 2nd order upwind flux evaluation method and a two-stage Runge-Kutta time advance scheme. For each interior volume cell, the net flux will be calculated for a specific time step using the second order upwind method. The flux for the face $i = 3/2$ is displayed in equation 1.3. The flux is evaluated at each face and then the faces surrounding a single cell are summed to find the flux through that cell. Integrating this into equation ?? the wave equation can be evaluated using equation 1.4

$$T_{\frac{3}{2}} = \frac{3\bar{T}_1 - \bar{T}_0}{2} \quad (1.3)$$

$$\frac{\partial T}{\partial t} + u \left(\frac{3\bar{T}_i - 4\bar{T}_{i-1} + \bar{T}_{i-2}}{2\Delta x} \right) = 0 \quad (1.4)$$

Boundary conditions will be implemented using ghost cells, allowing the interior scheme to remain the same during calculation of bordering cells. Because the interior scheme requires

two upwind values, two ghost cells are required for this scheme to run on the first interior cell $i = 1$. The ghost cells are calculated such the boundary condition is enforced at $i = 1/2$.

$$\bar{T}_0 = 2\bar{T}_w - \bar{T}_1 \quad (1.5)$$

$$\bar{T}_{-1} = -2\bar{T}_w + 3\bar{T}_0 \quad (1.6)$$

The value for the ghost cells implementing Neumann Boundary at $x = 0$ is calculated using equation 1.5 for cell $i = 0$ and 1.6 for cell $i = -1$.

2. IMPLEMENTATION

2.1. PROGRAM OVERVIEW

The C/C++ language was selected for this programming assignment. The scripts were compiled using g++/gcc version 5.4.0 on Ubuntu 16.04.02 and are available in the attached zip or for clone via the link provided: <https://github.com/j16out/cfd510>. The program itself is broken into 3 pieces and two levels to produce a modular set that makes it easier to apply to different problems. The highest level contains the "macro" or the main function which can be modified for different problems. The numerical directory contains numerical.cpp script and its header file numerical.hpp. This set contains all the functions necessary for solving the problem numerically. Table 2.1 displays all functions with a short description of each. The vroot directory contains the scripts necessary for drawing data. These scripts make use of the ROOT-v6 libraries. ROOT is a popular data analysis framework primarily written in C++ and Python. For more information on ROOT libraries visit the link provided: <https://root.cern.ch/>.

Function	Purpose
set_array_size()	Set size of array if not default(160x160)
set_ghostcells()	Set Ghost cells and update boundary conditions
set_zero()	Zero entire array including ghost cells
print_array()	Prints array in terminal style output
gs_iter_SOR()	Performs single Gauss-Seidel iteration with over-relaxation, over first in y then x
calc_source()	Calculates the source term for the problem
calc_newcell()	Calculates new solution at single grid point
get_surcells()	Gets current solution of neighboring cells
solve_arraySOR()	Solves for solution by performing many GS-iterations until defined convergence is met
get_discrete_Error()	Returns error for solution at $P(1/2,1/2)$ for Poisson Problem
get_solution()	Returns Solution at $P(1/2,1/2)$ for Poisson Problem via averaging scheme
get_l2norm()	Returns L^2 norm error between two arrays of equal size
set_analytic()	Set array values to predefined exact solution

Table 2.1: List of Program Functions

The functions act on a structure called *carray* which contains the solution domain array and its various parameters. The Struct contains the main array, its defined working area (mesh size), data storage vectors, iteration count, and the represented dimension between points. Having these organized in a struct provides a compact way of passing and modifying the array and all its pertinent parameters. The outline of the struct used for the wave equation problem is shown below.

```

struct carray{
//arrays
float mcellSOL [maxx][maxy]; //first stage and solution mesh
float mcellSOL2 [maxx][maxy]; //second stage solution mesh
float mcellFI [maxx][maxy]; //first stage flux
float mcellFI2 [maxx][maxy]; //second stage flux

//array attributes
int sizex = maxx;
int sizey = maxy;
float DIM1 = 0;

//current time
float tstep = 0;
float ctime = 0;

//temporary cells to store

```

```

float Tim1_j=0.0;
float Tim2_j=0.0;
float Ti_j=0.0;

};

```

2.2. 1D WAVE PROBLEM

The program was used to solve and find solutions to the 1D wave equation with information moving left to right (2.1) with satisfying conditions (2.2). An exact solution to this problem is provided and written as $T(x, t)$ and displayed in equation 2.3 which will enabled a comparison for solution error assessment.

$$\frac{\partial T}{\partial t} + u \frac{\partial T}{\partial x} = 0 \quad (2.1)$$

$$0 \leq x \leq x_{max} \quad 0 \leq t \quad u = 2 \quad (2.2)$$

$$T(x, t) = \sin(2\pi(2t - x)) \quad (2.3)$$

2.3. BOUNDARY CONDITIONS

Boundary Conditions are implemented through the calculation of the ghost cells. For the purely up wind scheme, we require two ghost cells to be calculated. The ghost cell is calculated such that the boundary is implemented at the face between the first ghost cell and first border cell. Below shows the function that implements boundaries for the Wave problem.

```

void set_ghostcells(carray & myarray)
{
//set boundary conditions in ghost cells
myarray.mcellSOL2[0][1] = -2.0*(sin(4.0*PI*myarray.ctime)) +
    3.0*myarray.mcellSOL[1][1];
myarray.mcellSOL2[1][1] = 2.0*(sin(4.0*PI*myarray.ctime)) -
    myarray.mcellSOL[2][1];
}

```

A loop is not necessary as this is purely a 1D problem therefore only two cells need to be updated each time step iteration. This subroutine will need to be modified when the program is applied to different problems or using different flux schemes at the boundary. At this stage is not changeable during run-time. Initial Conditions are set before the problem is solved using the subroutine shown below.

```

void set_intial_cond(carray & myarray)
{
float DIM1 = myarray.DIM1;
float dx =0.0;

```

```

float f;
for(int j = 1; j < myarray.sizey-1; ++j)
{
    for(int i = 2; i < myarray.sizey; ++i)
    {
        dx = (i-1.5)*DIM1;
        f = -sin(2.0*PI*dx);
        myarray.mcellSOL[i][j] = f;
    }
}

```

The loop calculates all the values of T at time $t=0$ for varying x using equation 2.4. Similar to the boundary conditions the initial conditions are not modifiable during run-time and will need to be tailored to specific problems.

$$T(x,0) = -\sin(2\pi x) \quad (2.4)$$

2.4. ITERATION SCHEME

A RK2 routine was implemented to solve the system of equations approximating the solution at each grid point which is denoted as function *solve_arrayRK2*. The routine starts from the first interior cell $i = 2$ and iterates flux integration schemes until $i = i_{max}$. Once the flux values have been gathered for each face and expressed as a single cell value a subroutine calculates the time advance solution at each cell. Below shows both the flux calculating and solution calculating subroutines respectively.

```

void get_FIarray(carray & myarray, int stage)
{
    for(int j = 1; j < myarray.sizey-1; ++j)
    {
        for(int i = 3; i < myarray.sizey; ++i)
        {
            //----get surrounding cells and compute new cell-----//
            get_surcells(myarray, i, j, stage);
            float newcell = calc_2nd_UW(myarray);
            //-----update current cell-----//
            if(stage == 1)
                myarray.mcellFI[i][j] = newcell;
            if(stage == 2)
                myarray.mcellFI2[i][j] = newcell;
        }
    }
}

void get_RK2(carray & myarray, int stage)
{

```

```

if(stage == 1)
{
for(int j = 1; j < myarray.sizey-1; ++j)
{
    for(int i = 2; i < myarray.sizez; ++i)
    {
        myarray.mcellSOL2[i][j] =
            myarray.mcellSOL[i][j]-myarray.tstep*(myarray.mcellFI[i][j]);
    }
}
}
if(stage == 2)
{
for(int j = 1; j < myarray.sizey-1; ++j)
{
    for(int i = 2; i < myarray.sizez; ++i)
    {
        myarray.mcellSOL2[i][j] =
            myarray.mcellSOL[i][j]-myarray.tstep*((myarray.mcellFI2[i][j]+myarray.mcellFI[i][j])/2.0);
    }
}
}
}

```

It should be noted the first interior cell $i = 2$ flux is calculated in a separate sub routine called *get_FIarray_1stcell*. An outer solving loop implements the RK2 scheme via two stages for all required time steps. After the two stages are computed the scheme is advanced by one time step and the boundary conditions are re-evaluated and implemented by calculating the new ghost cell values for the latest solution. Below displays the solution solving loop:

```

void solve_arrayRK2(carray & myarray, float tmax, float cfl)
{
float tstep = (cfl*(myarray.DIM1))/2.0;
myarray.tstep = tstep;
float ctime = myarray.ctime;

//set initial conditions
set_initial_cond(myarray);
set_ghostcells(myarray);

while(ctime < tmax-tstep)
{
    for(int h = 1; h <= 2; ++h)
    {
        get_FIarray_1stcell(myarray, h); //(array, stage)
        get_FIarray(myarray, h); //(array, stage)
        get_RK2(myarray, h); //(array, stage)
    }
}

```



```

//mv sol2 back to array sol1
mv_SOL2_to_SOL1(myarray);

//flux at boundary
set_ghostcells(myarray);//evaluates to ghost cells for array

myarray.ctime = myarray.ctime+myarray.tstep;
ctime = myarray.ctime;
++n;
}
}

```

The scheme iterates until the required end time is reached. The time steps are determined by the CFL number and the mesh size. Once the solution has been solved a analytical solution is produced and used for an error comparison.

2.5. VALIDATION

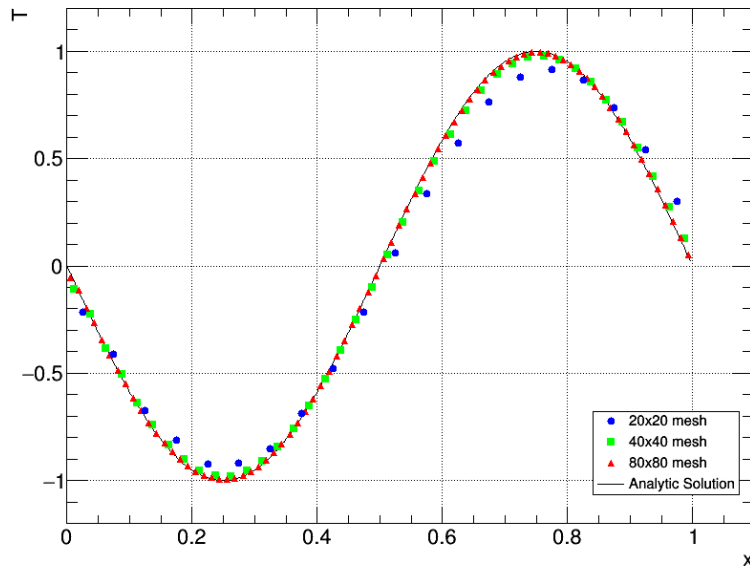


Figure 2.1: The maximum change in solution for $\omega = 1, 1.3, 1.5$ on a 20 x 20 mesh

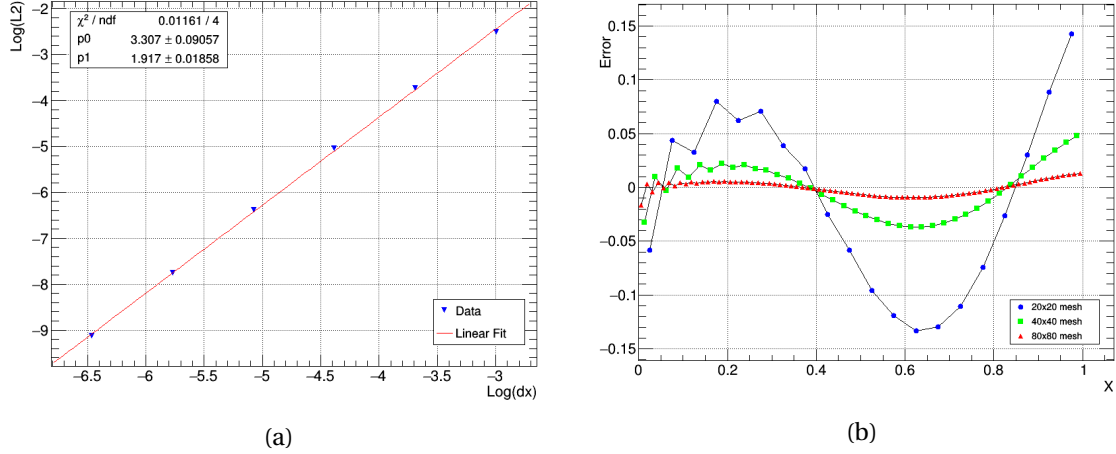


Figure 2.2: The Numerical Solution (a) and Solution Error (b) to Laplace Problem for 10x10 mesh

The full solution was numerically calculated and is display in figure 2.1 along with the solution error (computed - exact). The maximum change in solution and the L^2 norm as a function of iteration for a 10 x 10 mesh with $\omega = 1.0, 1.5$ are displayed in figure 2.3. The exact amount of iterations to solve this problem for a 10x10 mesh was found to be 137 and 57 for $\omega = 1$ and $\omega = 1.5$ respectively when maximum solution threshold was set to 10^{-7} .

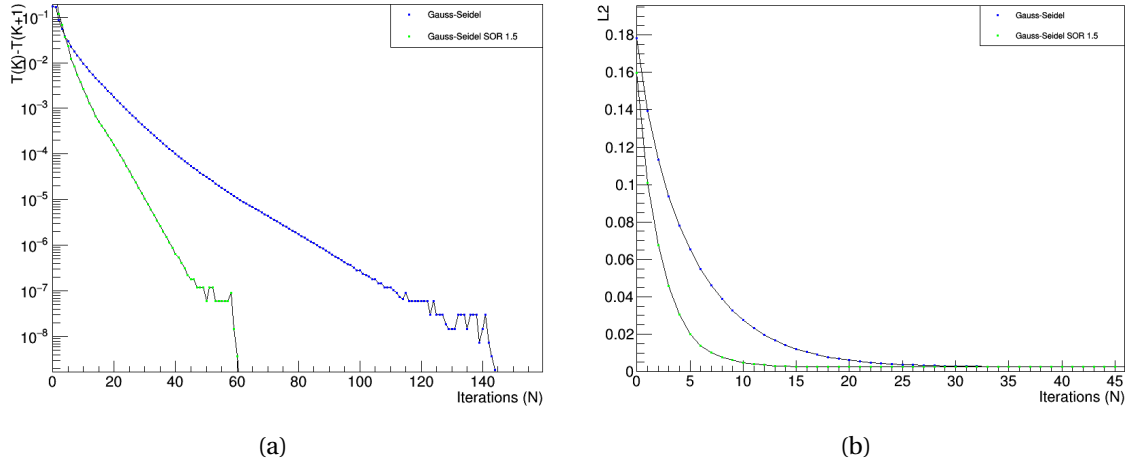


Figure 2.3: Maximum solution difference (a), and the L^2 norm of the error in converged solution (b) for $\omega = 1, 1.5$ for a 10x10 mesh

The L^2 norm was found to be 2.45×10^{-3} for both cases. It clear from the plots of both the error difference and the l2 norm that the method employing over-relaxation method drastically reduces the computation time for the same problem. Its also noticed the L^2 norm also converges faster with the over-relaxation method. One would assume higher values of ω

would result in faster and shorter iteration times however with ω values above 1.5 the scheme can become unstable. Essentially the change between the previous and current solution is too great and eventually the routine gets to a point where it constantly over shoots the convergence point without being able to reduce maximum solution change. In this case the ω value needs to be reduced in order to attain lower values for the solution change. The function `solve_arraySOR()` contains a small subroutine which checks for conditions that might indicate an unstable condition, and in the event reduces the ω value. For future operations a function should be written that decreases the ω as a function of its tolerance, as the current method is not sufficient for predicting all situations of instability.

2.6. CONVERGENCE BEHAVIOR AND ACCURACY

In order to gauge the convergence behavior of the problem the program was run until steady state for ω values of 1, 1.3, and 1.5. The maximum change in solution at each iteration was plotted as a function of iteration count displayed in figure 2.4. The total iterations were found to be 472, 310, and 226 for ω values of 1, 1.3, and 1.5 respectively. This data shows the effect of over-relaxation (SOR) on computing time. Although the speed of the program can vastly be increased this is only true for stable values of ω . When unstable a solution may never be reached within a defined maximum change in solution, and therefore would take many more iterations than stable schemes. Generally when $\omega > 1.5$ instability should be expected for the Gauss-Seidel with over-relaxation.

Mesh Size	L^2 norm	Δx	Order	Ratio
10 x 10	$2.457 \cdot 10^{-3}$	0.1	-	-
20 x 20	$6.384 \cdot 10^{-4}$	0.05	1.94	3.84
40 x 40	$1.603 \cdot 10^{-4}$	0.025	1.96	3.98
80 x 80	$3.827 \cdot 10^{-5}$	0.0125	2.00	4.18

Table 2.2: Table of L^2 norms for increasing mesh size

Tolerance	L^2 norm	ω
10^{-6}	$1.767 \cdot 10^{-4}$	1.0
10^{-7}	$1.605 \cdot 10^{-4}$	1.0
10^{-8}	$1.603 \cdot 10^{-4}$	1.0
10^{-9}	$1.603 \cdot 10^{-4}$	1.0
10^{-10}	$1.603 \cdot 10^{-5}$	1.0
10^{-7}	$1.603872 \cdot 10^{-5}$	1.2
10^{-7}	$1.603879 \cdot 10^{-5}$	1.4
10^{-7}	$1.603896 \cdot 10^{-5}$	1.6

Table 2.3: Table of L^2 norms in relation to tolerance and over-relaxation parameter ω

The accuracy of the program was estimated by calculating and tabulating the L^2 norms of solution error for multiple mesh sizes. Using table 2.2 we can estimate the order of error by plotting $\log(error)$ vs $\log(\Delta x)$ and calculating the slope. The slope and therefore the order of accuracy was estimated to be 2.003 which rounds to 2nd order accuracy. The ratio between the errors is also indicated in table 2.2, which gives us an indication of how the error is reduced relative to mesh refinement. In this case the effect of mesh refinement can be generalized in doubling the resolution we find the L^2 error is decrease by a quarter. Additionally as we refine the mesh se find that order converges on the order estimated by the selected estimation of solution fluxes which was 2nd order. The maximum solution change was lowered to 10^{-9} in order to ensure the error being measured was discretization error and not a lack of convergence. The error trend was also examined for the algorithm variables ω and the Convergence Tolerance which is displayed in table 2.3. This table shows reducing the tolerance generally reduces the discretization error.

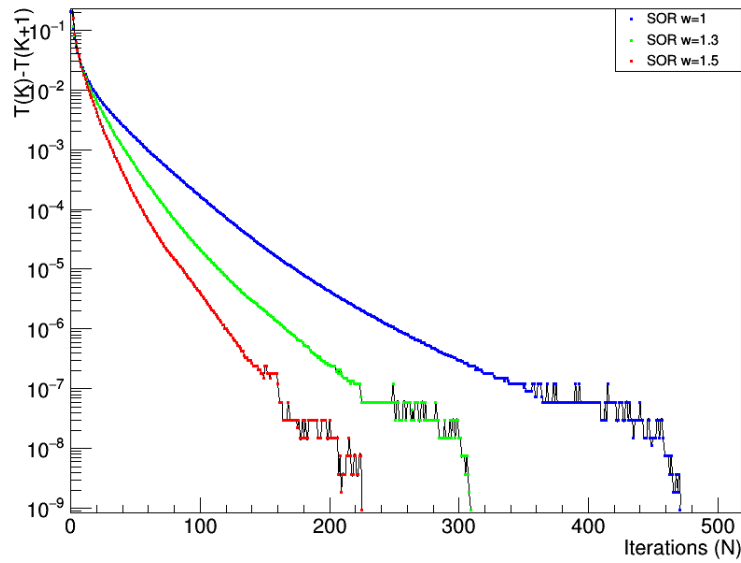


Figure 2.4: The maximum change in solution for $\omega = 1, 1.3, 1.5$ on a 20 x 20 mesh

3. POISSON PROBLEM

3.1. OVERVIEW

Calculating pressure for in-compressible flows is done by solving the Poisson problem for pressure. Starting with the momentum equations one can obtain a simplified expression for Poisson problem (equation 3.1). This equation allows us to simply add a source term (equation 3.2) to our existing Laplace code and compute the pressure for the previous square domain as

displayed in equation 1.4.

$$\frac{\partial^2 P}{\partial x^2} + \frac{\partial^2 P}{\partial y^2} = - \left(\left(\frac{\partial u}{\partial x} \right)^2 + 2 \frac{\partial v}{\partial x} \frac{\partial u}{\partial y} + \left(\frac{\partial v}{\partial y} \right)^2 \right) \quad (3.1)$$

$$S_{ij} = (3x^2 - 3y^2)^2 + 2(36x^2 y^2) + (3y^2 - 3x^2) \quad (3.2)$$

For boundary conditions $\frac{\partial P}{\partial n} = 0$ at $x = 0$ and $y = 0$, and $P = 5 - \frac{1}{2}(1 + y^2)^3$ for $x = 1$ and $P = 5 - \frac{1}{2}(1 + x^2)^3$ for $y = 1$ were implemented.

3.2. RESULTS

The Full solution to the Poisson problem was computed using $\omega = 1.3$ and is displayed in figure 3.1. This data was used to estimate the value for P at $(x, y) = (\frac{1}{2}, \frac{1}{2})$ from a 10x10 up to a 160x160 grid. The value of P was estimated using the average of the four surrounding cells. The solutions for P at $(x, y) = (\frac{1}{2}, \frac{1}{2})$ are tabulated in table 3.1 as P_1, P_2 , and P_3 for decreasing grid size. The solution was found to be 4.9385 with a GCI_{fine} uncertainty of %0.0043 in the fine-grid solution. The uncertainty was calculated using the procedure described in the ASME solution accuracy handout.[Celik, 2006] The local order of accuracy for $P(\frac{1}{2}, \frac{1}{2})$ was found on average to be 1.99 which sets the accuracy at 2nd order. Table 3.2 displays the accuracy behavior with mesh refinement for the Poisson problem.

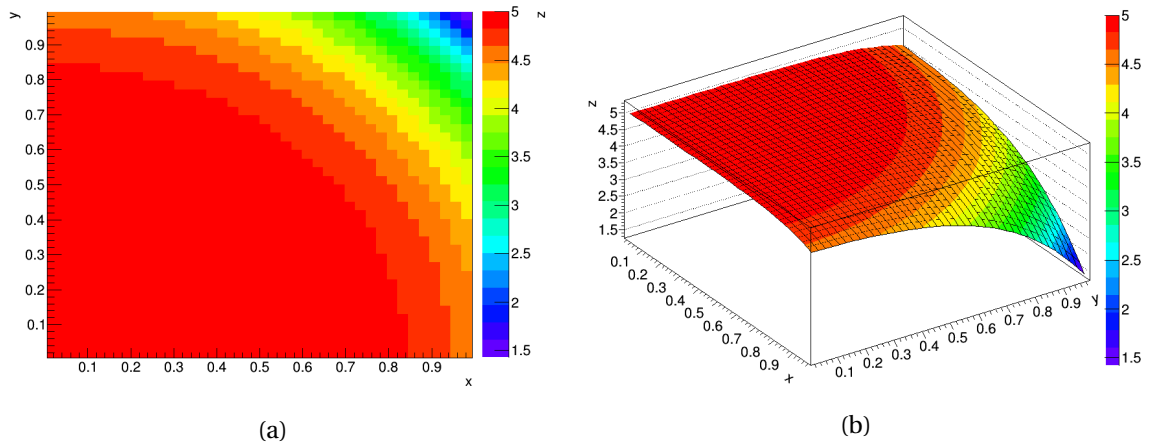


Figure 3.1: Calculated Solution of Poisson problem for 20 x 20 grid

Values	$P(\frac{1}{2}, \frac{1}{2})$
N_1, N_2, N_3	40, 20, 10
r_{21}	1.909091
r_{32}	1.833333
P_1	4.937872
P_2	4.938539
P_3	4.941434
p	2.468
P_{ext}^{21}	4.937703
e_a^{21}	%2.684
e_{ext}^{21}	%0.0034
GCI_{fine}^{21}	%0.0043

Table 3.1: Table of L^2 norms for increasing mesh size

Mesh Size	Solution	Apparent Order	Rel Error	Extrap Value	GCI
10 x 10	4.941434	-	-	-	-
20 x 20	4.938539	-	0.0058%	-	-
40 x 40	4.937872	2.468819	0.0037%	4.937703	0.0043%
80 x 80	4.938059	1.895416	0.0037%	4.937605	0.0019%
160 x 160	4.938802	1.624394	0.015%	4.936539	0.0315%

Table 3.2: Table of L^2 norms for increasing mesh size

Decreasing the grid size will help the solution converge closer to the exact values. If we image the grid size becoming infinitely small the grid would become continuous and essentially would describe the exact solution. However Decreasing grid size (increasing array size) has a hard hit on computational performance. Therefore before solving such a problem one should consider the exactness required for their problem and apply the appropriate mesh size. In the case where the mesh density can vary through out the problem domain it makes sense to refine the mesh in the areas of interest while providing a course mesh for areas that are understood relatively well. This will ensure the problem is solved the fastest while attaining the best approximate solution.

4. CONCLUSION

This project provides great inside to the internal algorithms used for calculating numerical solutions for symmetric grids. The Laplace problem provided a great platform for developing and testing the program. Having the analytic solution really help fine tune the program and helped given an idea on how accurate the solutions could be. Applying the program to

Poisson problem showed how one could find a solution and still estimate error without a known solution to compare with. Numerical methods provide a great way of solving difficult problems and until new methods are discovered for finding exact solution will remain the main method for finding solutions to unsolvable problems.

A. APPENDIX

A.1. POISSON.CPP

```
/*-----*/
Main Program for finding pressure for incompressible flows using Poisson
equations.
Finds solution at P(1/2,1/2) for Land discretization error for w values of 1
for
20x20,40x40 and 60x60 array

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compiled using g++/gcc version 5.4.0 on Ubuntu 16.04.02 and are available for
clone
via the link provided: url{https://github.com/j16out/
//-----*/

#include <vector>
#include <iostream>
#include <cstdlib>
#include <fstream>
#include <string>
#include <vector>
#include <algorithm>
#include <sstream>
#include <math.h>
#include "TApplication.h"
#include "vroot/root.hpp"
#include "numerical/numerical.hpp"

using namespace std;

#define E07 0.0000001
#define E08 0.00000001
#define E09 0.000000001
#define E10 0.0000000001
#define E11 0.00000000001
```

```

int main(int argc, char **argv)
{

carray poisson1;//my main array
carray poisson2;
carray poisson3;

//set array size or default used 162x162
set_array_size(poisson1, 20, 20, 1.0);//array, xsize, ysize, dimension
set_array_size(poisson2, 40, 40, 1.0);
set_array_size(poisson3, 80, 80, 1.0);

//set ghost cells as boundary conditions

set_zero(poisson1);
set_ghostcells(poisson1);//set ghost cells/boundaries
//print_array(poisson1);//print array in terminal

set_zero(poisson2);
set_ghostcells(poisson2);
//print_array(poisson2);

set_zero(poisson3);
set_ghostcells(poisson3);
//print_array(poisson3);

//-----GS SOR w=1.3 loop 1-----//

solve_arraySOR(poisson1, E11, 1.3);
cout << "Solution: " << get_solution(poisson1) << "\n";

//-----GS SOR w=1.3 loop 2-----//

solve_arraySOR(poisson2, E11, 1.3);
cout << "Solution: " << get_solution(poisson2) << "\n";

//-----GS SOR w=1 loop 3-----//

solve_arraySOR(poisson3, E11, 1.3);
cout << "Solution: " << get_solution(poisson3) << "\n";

//-----calc error based on ASME-----//

```



```

get_discrete_Error(poisson3, poisson2, poisson1, 1.0);

//-----Draw Data-----//

if(1)//start root application
{
    TApplication theApp("App", &argc, argv);
    draw_3DgraphP(poisson3);//draw 3d graph
    theApp.Run();
}

//end
}

```

A.2. NUMERICAL.HPP

```

#ifndef numerical_INCLUDED
#define numerical_INCLUDED

#include <vector>
#include <iostream>
#include <cstdlib>
#include <fstream>
#include <string>
#include <vector>
#include <algorithm>
#include <sstream>
#include <math.h>
#include <iomanip>

using namespace std;

#define BIG 1000000
#define maxx 160
#define maxy 160
#define PI 3.141592654

struct carray{

```

```

float mcell [maxx][maxy];
vector<float> l2norm;
vector<float> diff;
int sizex = maxx;
int sizey = maxy;
int iterations = 0;
float DIM1 = 0;

};

void set_array_size(carray & myarray, int x, int y, float DIM); //set array size

void set_ghostcells(carray & myarray); //set ghost cells

void set_zero(carray & myarray); //zero entire array

void print_array(carray & myarray); //print array in terminal

float gs_iter_SOR(carray & myarray, float omega); //complete one gauss-seidel
iteration with SOR

float calc_source(carray & myarray, int i, int j); //calculate source term

float calc_newcell(carray & myarray, float source, float Tip1_j, float Tim1_j,
float Ti_jp1, float Ti_jm1); //calculate new cell value based on 2nd order
scheme

void get_surcells(carray & myarray, float & Tip1_j, float & Tim1_j, float &
Ti_jp1, float & Ti_jm1, int i, int j); //obtain values of surrounding cells

void solve_arraySOR(carray & myarray, float E0, float w); //solve the array
using gs-iterations

void get_discrete_Error(carray ray1, carray ray2, carray ray3, float
DIM); //get error using 3 arrays based on ASME solution accuracy handout

float get_solution(carray & myarray); //find solution at  $p(1/2, 1/2)$  for Poisson

float get_l2norm(carray & myarray, carray myarray2); //get estimated vale for
l2 norm between arrays

#endif

```

A.3. NUMERICAL.CPP

```

#include "numerical.hpp"

//-----set array size (working area excluding ghost)-----//

void set_array_size(carray & myarray, int x, int y, float DIM)
{
    if(x < 160 && y < 160)
    {
        myarray.size_x = x+2;
        myarray.size_y = y+2;
        myarray.DIM1 = DIM/(x);
    }
    else
        cout << "Array size too big, setting to default 160" << "\n";
}

//-----zero array-----//

void set_zero(carray & myarray)
{
    for(int i = 0; i < myarray.size_x; ++i)
    {
        for(int j = 0; j < myarray.size_y; ++j)
        {
            myarray.mcell[i][j] = 0; //set everything to zero
        }
    }
}

//-----set ghost cells for Poisson-----//

void set_ghostcells(carray & myarray)
{
    float DIM1 = myarray.DIM1;

    //set boundary conditions in ghost cells
    for(int i = 1; i < myarray.size_x-1; ++i)
        {float d = (i-0.5)*DIM1;

```

```

//neumann boundaries
myarray.mcell[i][0] = myarray.mcell[i][1]; //top ghost cells
myarray.mcell[0][i] = myarray.mcell[1][i]; //left ghost cells

//dirichlet boundaries
myarray.mcell[i][myarray.size_x-1] = 2.0*(5.0-((1.0/2.0)*pow((1.0+pow(d,
    2)),3))) - myarray.mcell[i][myarray.size_x-2];
myarray.mcell[myarray.size_y-1][i] = 2.0*(5.0-((1.0/2.0)*pow((1.0+pow(d,
    2)),3))) - myarray.mcell[myarray.size_x-2][i];

    }
}

//-----Guass-Seidel SOR-----//

float gs_iter_SOR(carray & myarray, float omega)
{
    float DIM1 = myarray.DIM1; //get dimensions of array (not grid size)
    float Tip1_j, Tim1_j, Ti_jp1, Ti_jm1; //define values for surrounding cells
    float l2sum = 0.0;
    float sx = myarray.size_x-2;
    float sy = myarray.size_y-2;

    float dx = 0.0; //change in x
    float dy = 0.0; //change in y

    carray oldarray=myarray;

    //-----iterate through all x for steps y -----//
    set_ghostcells(myarray);

    for(int j = 1; j < myarray.size_y-1; ++j)
    {

        for(int i = 1; i < myarray.size_x-1; ++i)
        {

            dx = (i-0.5)*DIM1;
            dy = (j-0.5)*DIM1;

            //----get surrounding cells and compute new cell-----//
            get_surcells(myarray, Tip1_j, Tim1_j, Ti_jp1 , Ti_jm1, i, j);
            float source = calc_source(myarray, i, j);
            float newcell = calc_newcell(myarray, source, Tip1_j, Tim1_j, Ti_jp1 ,
                Ti_jm1);

```

```

    //----apply over-relaxation-----//
    float delta = newcell - myarray.mcell[i][j];
    float newcellSOR = myarray.mcell[i][j] + omega*(delta);

    //-----update current cell-----//
    myarray.mcell[i][j] = newcellSOR;

}

}

set_ghostcells(myarray);
//-----iterate through all y for steps x -----//
for(int i = 1; i < myarray.sizey-1; ++i)
{

    for(int j = 1; j < myarray.sizey-1; ++j)
    {

        dx = (i-0.5)*DIM1;
        dy = (j-0.5)*DIM1;

        //----get surrounding cells and compute new cell-----//
        get_surcells(myarray, Tip1_j, Tim1_j, Ti_jp1 , Ti_jm1, i, j);
        float source = calc_source(myarray, i, j);
        float newcell = calc_newcell(myarray, source, Tip1_j, Tim1_j, Ti_jp1 ,
            Ti_jm1);

        //----apply over-relaxation-----//
        float delta = newcell - myarray.mcell[i][j];
        float newcellSOR = myarray.mcell[i][j] + omega*(delta);

        //-----update current cell-----//
        myarray.mcell[i][j] = newcellSOR;

    }

}

float maxdiff = -1.0;

for(int i = 2; i < myarray.sizey-2; ++i)
{
    for(int j = 2; j < myarray.sizey-2; ++j)

```

```

    {
        float diff = abs(oldarray.mcell[i][j] - myarray.mcell[i][j]);
        if(diff > maxdiff)
        {
            maxdiff = diff;
            //cout << "coord " << maxdiff << " " << i << " " << j << "\n";
        }
    }
}

//for obtaining l2norm convergence

/*carray analytic;
set_array_size(analytic, 10, 10, 1.0);
set_analytic(analytic);
//float norm = get_l2norm(myarray, analytic);

myarray.l2norm.push_back(norm); */
myarray.diff.push_back(maxdiff);
++myarray.iterations;

return maxdiff;
}
//-----Get L2 norm for unknown
analytical-----//

float get_l2norm(carray & myarray, carray myarray2)
{
    float l2sum =0;
    float sx = myarray.sizeX-2;
    float sy = myarray.sizeY-2;

    for(int j = 1; j < myarray.sizeY-1; ++j)
    {
        for(int i = 1; i < myarray.sizeX-1; ++i)
        {

            float P = myarray.mcell[i][j];
            float T = myarray2.mcell[i][j];
            l2sum = l2sum + pow((P-T),2);

        }
    }

}

float l2 = sqrt(l2sum/(sx*sy));
cout << "L2 norm: " << l2 << "\n";
return l2;

```

```

}

//-----Solve array using
//GS-iterations-----//

void solve_arraySOR(carray & myarray, float E0, float w)
{
printf("\n\nSolving Grid size: %d Relaxation: %f\n", myarray.size_x, w);
bool relax_on = true;
float diff = 1; // current difference
float ldiff = BIG; // previous difference
int div = 0;
int update = 0;
int update2 = 100;

while(diff >= E0)
{
diff = gs_iter_SOR(myarray, w);

    if(diff > BIG) //avoid infinite loops if diverges
        break;

    if(update >= update2) //report difference every 100 steps
    {cout << "Update: step " << update << " Solution Change: " <<
        setprecision(9) << fixed << diff << " \n";
    //print_array(myarray);
        update2 = update2 + 100;
    }

    if(ldiff == diff) //checks for repeated values indication of instability for
        high w
        ++div;
    else
        div = 0;

    if(div > 3 && w > 1.1) //reduces over-relaxation for high w when unstable
    {
        w = 1.0;
        cout << "Relaxation Reduced to "<<w<<" @ " << myarray.iterations << " \n";
    }

    ldiff = diff;
    ++update;
}

cout << "Iterations: " << myarray.iterations << "\n";
}

```

```

//-----Print array in
terminal-----//

void print_array(carray & myarray)
{
cout << "\n";

    for(int j = 0; j < myarray.sizey; ++j)
    {
        cout << "\n|";
        for(int i = 0; i < myarray.sizez; ++i)
        {
            if(myarray.mcell[i][j] >= 0)
                cout << setprecision(3) << fixed << myarray.mcell[i][j] << "|";
            if(myarray.mcell[i][j] < 0)
                cout << setprecision(2) << fixed << myarray.mcell[i][j] << "|";
        }

    }
    cout << "\n";
}

//-----Calculate new cell value from neighbors
-----//

float calc_newcell(carray & myarray, float source, float Tip1_j, float Tim1_j,
    float Ti_jp1 ,float Ti_jm1)
{
    float DIM1 = myarray.DIM1;
    float chx = DIM1;
    float chy = DIM1;
    float temp = (pow(chx,2)*pow(chy,2)) / (2*(pow(chx,2)+pow(chy,2)));
    float newcell = (( (Tip1_j+Tim1_j)/pow(chx,2)) + ((Ti_jp1+Ti_jm1)/pow(chy,2))
        - source ) * temp ;

    return newcell;
}

//-----Get source term for poisson
problem-----//
float calc_source(carray & myarray, int i, int j)
{
    float DIM1 = myarray.DIM1;
    float dx = (i-0.5)*DIM1;
    float dy = (j-0.5)*DIM1;
    float source =
        -1.0*(pow(3*pow(dx,2)-3*pow(dy,2),2)+72.0*(pow(dx,2)*pow(dy,2))+pow(3*pow(dy,2)-3.0*pow(dx,2)
//source = 0 for Laplace problem

```



```

return source;
}

//-----Get average solution at point
(1/2)(1/2)-----//

float get_solution(carray & myarray)
{
float DIM1 = myarray.DIM1;
int sx = (myarray.sizeX)/2.0;
int sy = (myarray.sizeY)/2.0;
float sol =
(myarray.mcell[sx-1][sy]+myarray.mcell[sx][sy]+myarray.mcell[sx][sy-1]+myarray.mcell[sx-1][sy-1])/4.0;

printf("cell 1: %f cell 2: %f cell 3: %f cell 4: %f\n",
myarray.mcell[sx-1][sy],myarray.mcell[sx][sy],myarray.mcell[sx][sy-1],myarray.mcell[sx-1][sy-1]);
//for Poisson problem only, finds value based on average of four surrounding
cells

return sol;
}

//-----Get current cell
values-----//

void get_surcells(carray & myarray, float & Tip1_j, float & Tim1_j, float &
Ti_jp1 ,float & Ti_jm1, int i, int j)
{
float fcon = false;
float sizeX = myarray.sizeX;
float sizeY = myarray.sizeY;

Tip1_j = myarray.mcell[i+1][j];
Tim1_j = myarray.mcell[i-1][j];
Ti_jp1 = myarray.mcell[i][j+1];
Ti_jm1 = myarray.mcell[i][j-1];

}

//-----Get discrete error-----//

void get_discrete_Error(carray ray1, carray ray2, carray ray3, float DIM)
{
//Calculating error as described in paper "procedure for estimation and
reporting of uncertainty due to discretization in CFD applications"//

printf("\nCalculating Error...\n");
}

```

```

float h1 = DIM/ray1.size_x;
float h2 = DIM/ray2.size_x;
float h3 = DIM/ray3.size_x;

float sol1 = get_solution(ray1);
float sol2 = get_solution(ray2);
float sol3 = get_solution(ray3);

printf("h1: %f \nh2: %f \nh3: %f, \nsol1: %f \nsol2: %f \nsol3: %f\n",h1, h2,
      h3, sol1, sol2, sol3);

float r21 = h2/h1;
float r32 = h3/h2;

printf("\nr32: %f \nr21: %f\n",r32, r21);

float e32 = sol3-sol2;
float e21 = sol2-sol1;

float s = (e32/e21);
if(s >= 0)
s = 1;
else
s = -1;

float p_n = 0;
float p = (1/log(r21))*(abs(log(abs(e32/e21)))+0));

printf("intial guess: %f \n", p);

float diff = 1;

while(diff > 0.0000001)
{

float p_n =
(1/log(r21))*(abs(log(abs(e32/e21)))+log((pow(r21,p)-s)/(pow(r32,p)-s))
));
diff = abs(p_n -p);
//printf("p_n: %f p: %f diff: %f\n",p_n, p, diff);

p = p_n;
}

//
float sol_ext21 = (pow(r21, p)*sol1-sol2)/(pow(r21,p)-1.0);

```

```

float sol_ext32 = (pow(r32, p)*sol2-sol3)/(pow(r32,p)-1.0);

printf("order: %f \nphi_ext21: %f \nphi_ext32 %f\n",p, sol_ext21, sol_ext32);

float ea21 = abs((sol1-sol2)/sol1);

float e_ext21 = abs((sol_ext21-sol1)/sol_ext21);

float GCI_21 = (1.25*ea21)/(pow(r21,p)-1.0);

printf("ea21: %f \ne_ext21: %f \nGCI21 %f \n", ea21, e_ext21, GCI_21);
}

```

REFERENCES

- [Celik, 2006] Ismail B. Celik¹, Urmila Ghia, Patrick J. Roache and Christopher J. Freitas "Procedure for Estimation and Reporting Uncertainty Due to Discretization in CFD applications", West Virginia University, Morgantown WV, USA