

AMS 530 Final Project Parallel 2D Poisson Equation Solver

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Introduction and Motivation

- Physical laws often take the form of partial differential equations (PDEs), e.g., the Hamilton-Jacobi-Bellman (HJB) equation in control theory and the Schrodinger equation in quantum mechanics.
- There are various traditional numerical methods for solving PDEs; for example, the Finite Difference Method. However, when the grid size is large, it can be computationally intractable. Therefore, it is imperative to parallelize the algorithm to distribute the work to multiple processors.



Problem Description

Consider a 2D Poisson's equation with Dirichlet boundary conditions:

$$\Delta u(x,y) = f(x,y),$$
 in $\Omega := (0,1) \times (0,1),$ $u(x,y) = g(x,y),$ on $\partial \Omega,$

where Δ is the Laplacian operator, $f(x) = \sin(x - y)$, and g(x, y) = 1.

In this report, the PDE above will be solved by finite difference in parallel on three different grids, 100×100 , 200×200 , and 400×400 , using P = 2, 10, 20 processors.

Sequential FDM: Discretization

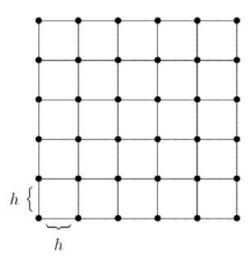
Discretization:

The unit square domain can be discretized as an $N \times N$ matrix:

$$x_i = i \cdot h, \ y_j = j \cdot h, \ i, j \in [0, N],$$

where the step size $h = dx = dy = \frac{1}{N}$, and

$$u_{i,j} = u(x_i, y_j), f_{i,j} = f(x_i, y_j) = sin(x_i - y_j).$$



- We wish to find the values of u(x, y) on this $N \times N$ grid.
- Values on the boundary, i.e., 4 edges of the square is known. Therefore, there are $(N-2) \times (N-2)$ unknown grid values.

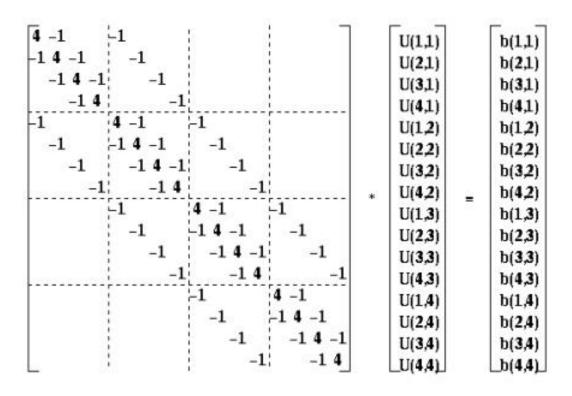


Sequential FDM: Discretization

By central difference approximation, for an interior grid point:

$$\frac{u_{i+1,j} + u_{i-1,j} - 2u_{i,j}}{h^2} + \frac{u_{i,j+1} + u_{i,j-1} - 2u_{i,j}}{h^2} = \sin(x_i - y_j)$$

Overall, we have a linear system of $(N - 2)^2$ equations with $(N - 2)^2$ unknown variables, which can be written in the form of Au = b as shown on the right.

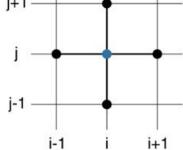


Sequential FDM: Five-point Stencil

Since this system of linear equations is sparse and strictly diagonally dominant, Jacobi method [4] can be applied to solve this system of equations. The updates for interior grid points $(u_{i,j}, i, j \in [1, N-1])$ are:

$$u_{i,j}^{n+1} = \frac{1}{4} (u_{i+1,j}^n + u_{i-1,j}^n + u_{i,j+1}^n + u_{i,j-1}^n - f_{i,j} \cdot h^2).$$

$$\downarrow +1 \qquad \qquad \downarrow \qquad \qquad \downarrow \qquad \qquad (2.1)$$



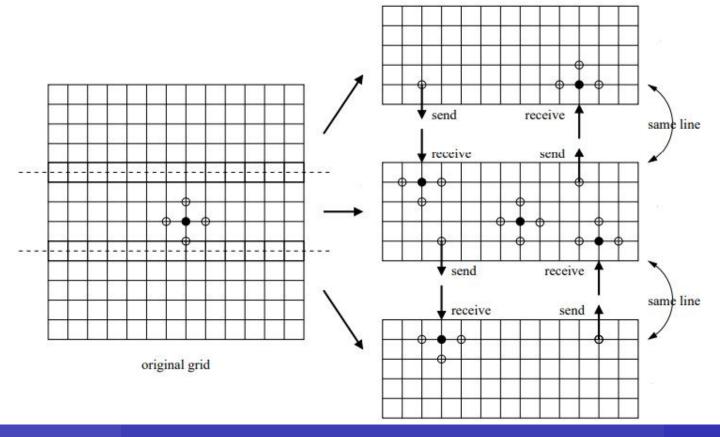
Start with an initial guess and keep applying Jacobi updates, the grid values will eventually converge to the solution of the linear system.

- 1. For the update of an interior grid point, we need the values from its 4 neighbours.
- 2. Each grid point's update is independent from other points' updates.



Parallelization: Domain Decomposition

■ In this report, we are going to parallelize the Jacobi method using so called "domain decomposition". The domain is decomposed into P "stripes".





Parallelization: Implementation

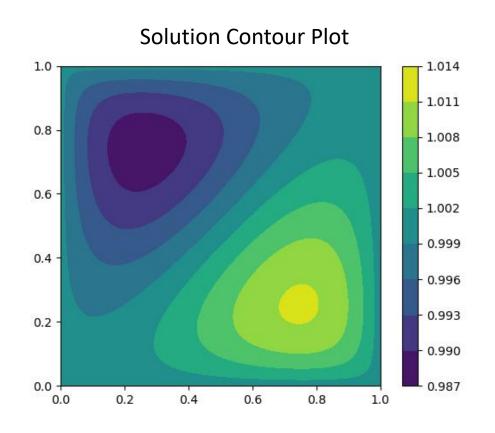
■ In a nutshell, in the parallel Jacobi algorithm, we first scatter the solution matrix u(x, y), and then there are two steps in each iteration: Information Exchange + Local Jacobi Updates. Finally, when the convergence is reached, we gather the solution matrix to the root processor.

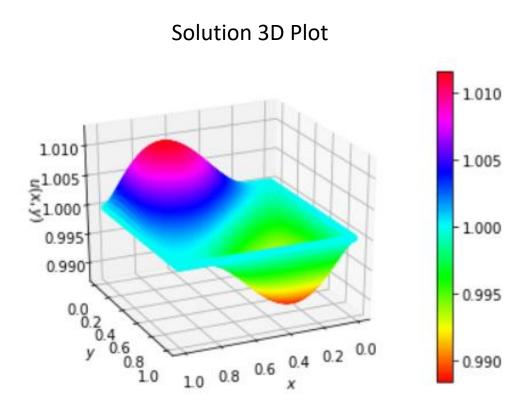
Information Exchange

Jacobi Updates

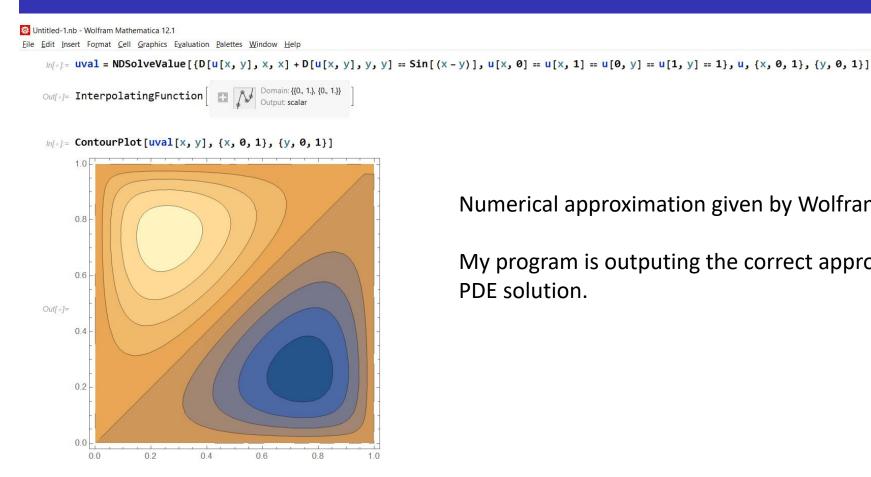
```
# exchange information, stripe partition
                                                      ldef jacobi(grid, f, dx, rank, size):
                                                          newgrid = np.zeros(shape=grid.shape)
def exchange info (rank, size, comm, sub u):
                                                          # updating non-bdy points
    if rank == 0:
                                                          if rank == 0:
         comm.Send(sub u[-2], dest=rank + 1)
                                                              newgrid[1:-1, 1:-1] = 0.25 * (grid[1:-1, :-2] + grid[1:-1, 2:] +
                                                                                     grid[:-2, 1:-1] + grid[2:, 1:-1] - f[1:, 1:-1] * dx ** 2)
         comm.Recv(sub u[-1], source=rank + 1)
                                                          elif rank == size -1:
    elif rank == size - 1:
                                                              newgrid[1:-1, 1:-1] = 0.25 * (grid[1:-1, :-2] + grid[1:-1, 2:] +
         comm.Send(sub u[1], dest=rank - 1)
                                                                                     grid[:-2, 1:-1] + grid[2:, 1:-1] - f[0:-1, 1:-1] * dx ** 2)
         comm.Recv(sub u[0], source=rank - 1)
                                                          else:
                                                              newgrid[1:-1, 1:-1] = 0.25 * (grid[1:-1, :-2] + grid[1:-1, 2:] +
    else:
                                                                                     grid[:-2, 1:-1] + grid[2:, 1:-1] - f[:, 1:-1] * dx ** 2)
         comm.Send(sub u[-2], dest=rank + 1)
         comm.Recv(sub u[-1], source=rank + 1)
                                                          # bdy points unchanged
                                                          newgrid[0, :] = grid[0, :]
         comm.Send(sub u[1], dest=rank - 1)
                                                          newgrid[-1, :] = grid[-1, :]
         comm.Recv(sub u[0], source=rank - 1)
                                                          newgrid[:, 0] = grid[:, 0]
    comm.Barrier()
                                                          newgrid[:, -1] = grid[:, -1]
                                                          return newgrid
```

Results: Solution Plots





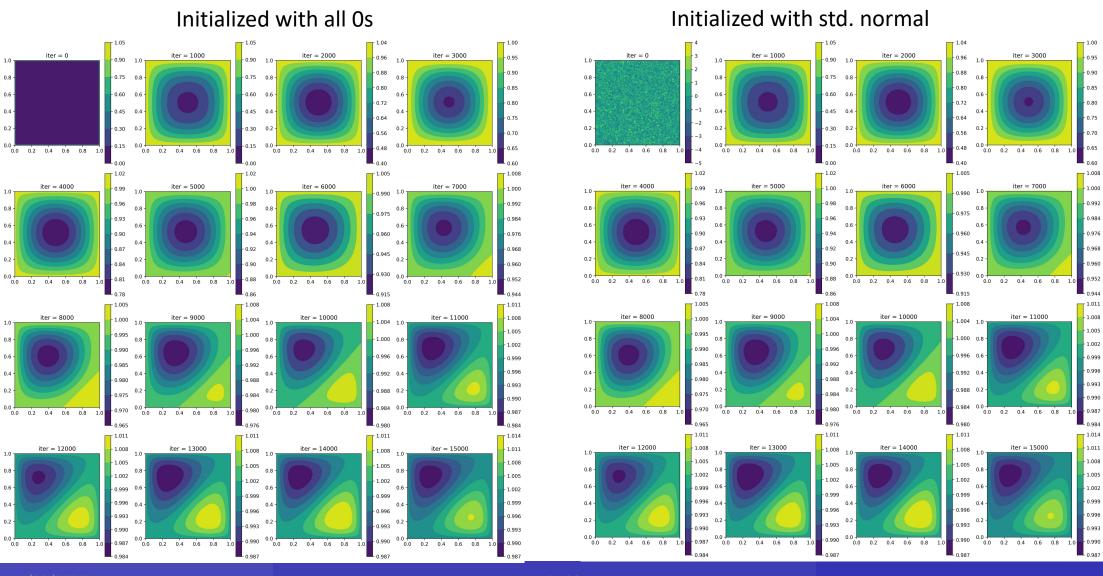
Results: Correctness Verification



Numerical approximation given by Wolfram

My program is outputing the correct approximation to the PDE solution.

Results: Convergence of Jacobi Method



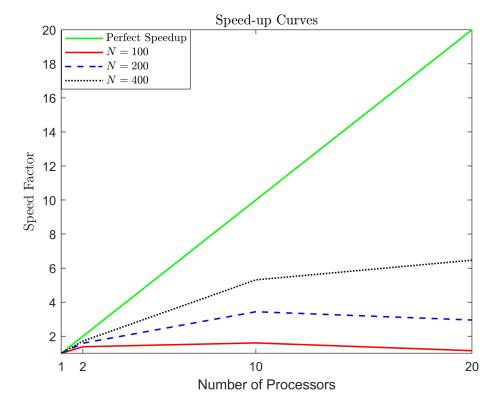
Results: Parallel Speed-up Results

	P = 1	P=2	P = 10	P = 20
N = 100	2.679693	1.934774	1.666728	2.326345
N = 200	28.899304	18.155115	8.402078	9.783527
N = 400	354.117078	206.414760	66.682876	54.755312

Table 1: Running Time in Seconds

	P=2	P = 10	P = 20
N = 100	1.385	1.607	1.152
N = 200	1.592	3.440	2.954
N = 400	1.716	5.310	6.467

Table 2: Speed-up Factors
$$S(P, N) = \frac{T(1,N)}{T(P,N)}$$



The speed-up gets better when the scale gets larger; this happens because the communication cost becomes less significant compared to the computation cost. Therefore, my program is efficient and scalable. We can expect almost a linear speed up asymptotically when N goes to infinity.



Thank you