Introduction to Project 3 Nonlinear PDE

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Overview

- Overview of Project 3
- Short review of the code
- Running the code and visualizing the output



The HPC PDE Application

The code solves a reaction diffusion equation known as Fischer's Equation

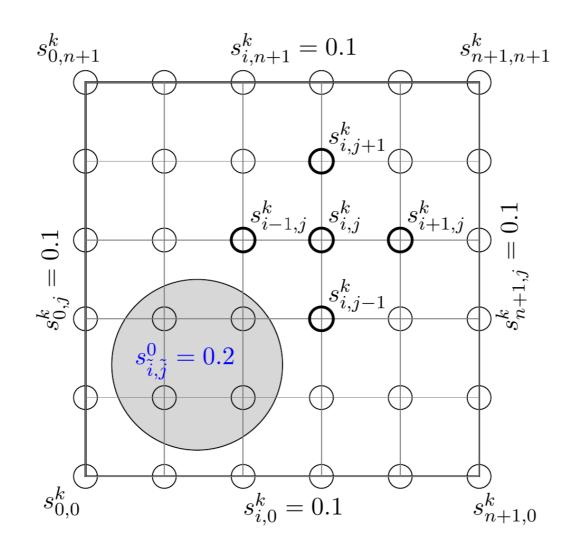
$$\frac{\partial s}{\partial t} = D(\frac{\partial^2 s}{\partial x^2} + \frac{\partial^2 s}{\partial y^2}) + Rs(1 - s)$$

Used to simulate travelling waves and simple population dynamics

- The species s diffuses
- And the population grows to a maximum of s=1

Initial and boundary conditions

- set the domain to be the unit square, i.e. $\Omega = (0,1)^2$
- fixed boundary values with $s(x) = 0.1 \ for \ x \in \partial \Omega$
- Initial values s^{init} are set to 0.1 except for a circular region with initial values of 0.2 in the lower left corner



The rectangular domain is discretized with a grid of dimension (n+2)x(n+2) points

 A second order finite difference discretization gives the following approximation for the spatial derivatives

$$\left(\frac{\partial^2 s}{\partial x^2} + \frac{\partial^2 s}{\partial y^2}\right)_{i,j} \approx \frac{1}{\Delta x^2} \left(-4s_{i,j} + s_{i-1,j} + s_{i+1,j} + s_{i,j-1} + s_{i,j+1}\right)$$

 A first order finite difference discretization is used for the approximation of the temporal derivative

$$(\frac{\partial s}{\partial t})_{i,j}^k \approx \frac{1}{\Delta t} (s_{i,j}^k - s_{i,j}^{k-1}),$$

Putting these together one obtains

$$\frac{1}{\Delta t}(s_{i,j}^k - s_{i,j}^{k-1}) = \frac{D}{\Delta x^2}(-4s_{i,j}^k + s_{i-1,j}^k + s_{i+1,j}^k + s_{i,j-1}^k + s_{i,j+1}^k) + Rs_{i,j}^k(1 - s_{i,j}^k)$$

Reformulate problem as

$$f_{i,j}^k := \left[-(4+\alpha)s_{i,j} + s_{i-1,j} + s_{i+1,j} + s_{i,j-1} + s_{i,j+1} + \beta s_{i,j} (1-s_{i,j}) \right]^k + \alpha s_{i,j}^{k-1} = 0$$

$$\alpha := \Delta x^2 / (D\Delta t)$$

$$\beta := R\Delta x^2/D$$

 One nonlinear equation for each grid point together they form a system of N=n*n equations Solve with Newton's method

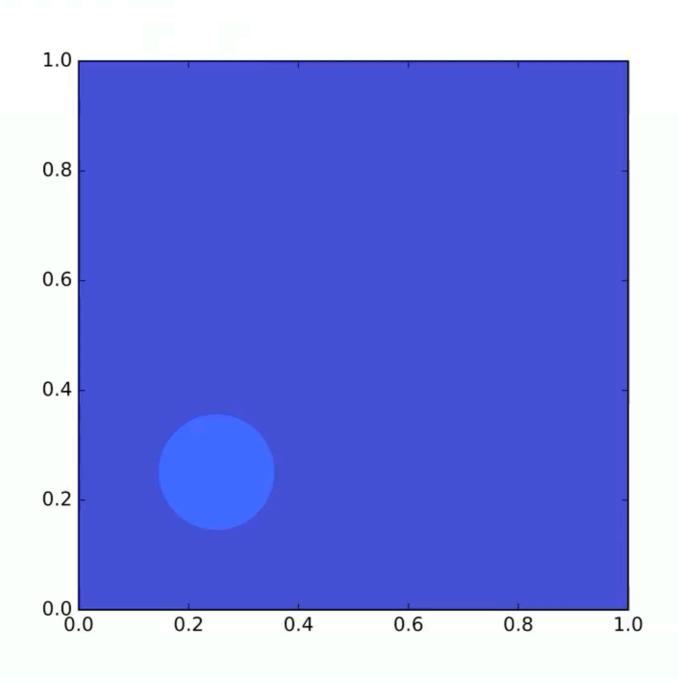
$$y^{l+1} = y^l - [J_f(y^l)]^{-1} f(y^l),$$

$$y^0 := s^{k-1}$$

• Each iteration of Newton's method has to solve a linear system Solve with matrix-free Conjugate Gradient solver

$$[J_f(y^l)]\delta y^{l+1} = f(y^l)$$
$$y^{l+1} = y^l - \delta y^{l+1}$$

Time Evolution of the Solution





- Most of the code is already implemented
- The focus is on the parallelisation of the code using OpenMP
- So let's look a little closer at each part of the code

Code Walkthrough

There are three modules of interest

- main.cpp: initialization and main time stepping loop
- linalg.cpp: the BLAS level 1 (vector-vector) kernels and conjugate gradient solver
- operators.cpp: the stencil operator for the finite difference discretization

the vector-vector kernels and diffusion operator are the only kernels that have to be parallelized



Linear algebra: linalg.cpp

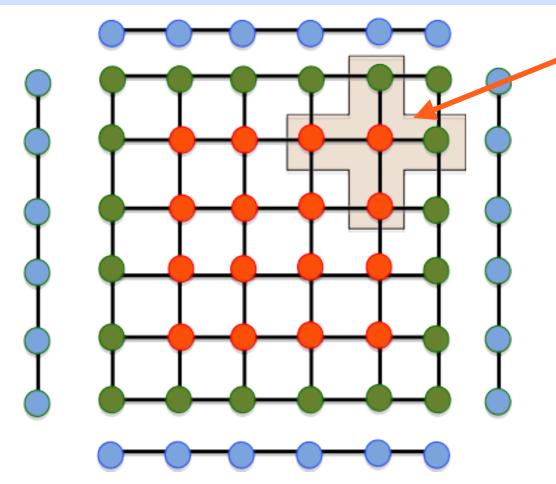
This file defines simple kernels for operating on 1D vectors, including

```
• dot product : x \cdot y : hpc_dot()
linear combination : z = \alpha x + \beta y where \alpha, \beta \in \mathbb{R}: hpc_lcomb()
```

The kernels of interest start with hpc_xxxxx()
 hpc = HPC Lab for CSE

For each parallelization approach that we will see (OpenMP, and later on MPI), each of these kernels will have to be considered.

Boundary-free stencil (interior stencil)



Interior Stencils have all points on interior grid

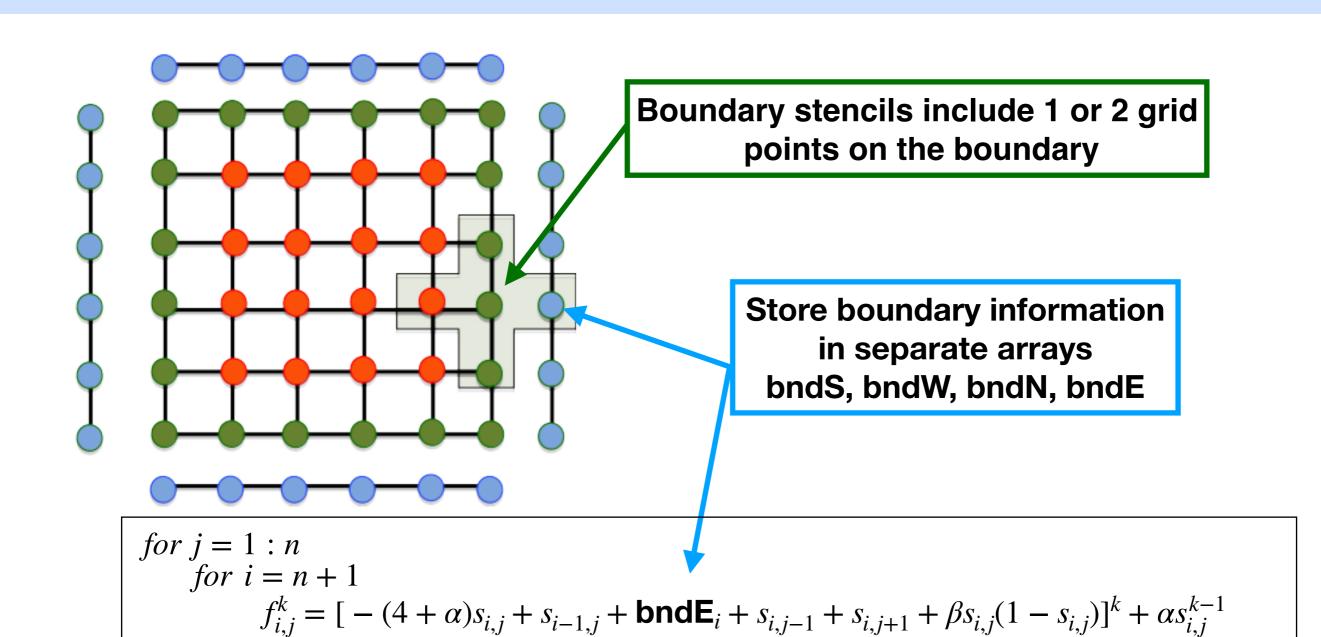
for
$$j = 1:n$$

for $i = 1:n$

$$f_{i,j}^k = [-(4+\alpha)s_{i,j} + s_{i-1,j} + s_{i+1,j} + s_{i,j-1} + s_{i,j+1} + \beta s_{i,j}(1-s_{i,j})]^k + \alpha s_{i,j}^{k-1}$$
end
end



Boundary stencil





end

end

Testing the code

- Get the code from Moodle or Github
- Compile and run

```
[user@eu-login]$ make
[user@eu-login]$ export OMP_NUM_THREADS=1
[user@eu-login]$ bsub -n 1 -W 00:10 -R "span[ptile=1]"
./main 128 100 0.005
```

Output

```
Welcome to mini-stencil!
version :: C++ serial
mesh :: 128 * 128 dx = 0.00787402
time :: 100 time steps from 0 .. 0.005
step 1 required 4 iterations for residual 7.21951e-07
step 2 required 4 iterations for residual 7.9975e-07
step 99 required 12 iterations for residual 9.36586e-07
step 100 required 12 iterations for residual 9.44772e-07
simulation took 1.58408 seconds
2853 conjugate gradient iterations, at rate 6341.35 iters/second
492 newton iterations
Goodbye!
```



Thank you for your attention and have fun with the Project!