# Introduction to the summer school mini-app code

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#### Overview

- What is a mini-app?
- An overview of our mini-app
- First look at the code.
- Run the code and visualize output



## **HPC Mini-Apps**

- Full HPC applications have complicated behavior
  - difficult to model or understand performance behavior
- Mini applications (mini-apps) are smaller codes that aim to characterize larger applications
  - typically a few hundred to a few thousand lines of code
- Are simpler to test and understand than full applications
- Used to test different hardware and programming languages
- Good for learning new techniques!



## Our Mini-App

- Throughout the summer school we will be using a mini-app to reinforce the lessons
  - During talks there will be small programming exercises to test out what you learn
  - Then you will get the opportunity to apply the techniques to the mini-app
- We will start with a serial version that has no parallel optimizations
  - By the end of the course we will have several different versions, one for each technique





# The Application

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

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

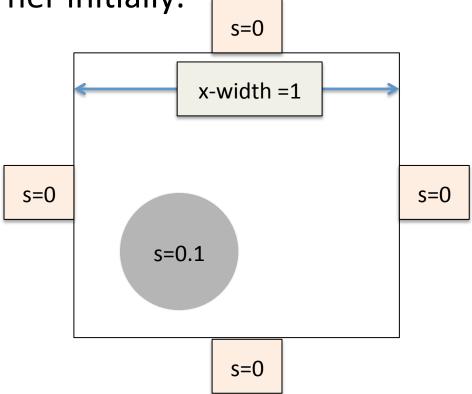
- Used to simulate travelling waves and simple population dynamics
  - The species s diffuses
  - And the population grows to a maximum of s=1





# Initial and boundary conditions

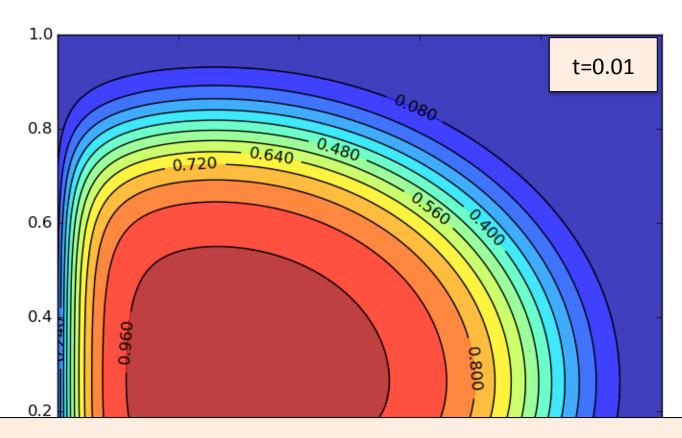
 The domain is rectangular, with fixed value of s=0 on each boundary, and a circular region of s=0.1 in the lower left corner initially.







## Time Evolution of Solution



For most cases we will run the solution until t=0.01.

- Long enough for something interesting to happen
  - Will clearly show if there is a problem



#### **Numerical Solution**

- The rectangular domain is discretized with a grid of dimension nx\*ny points
- A finite volume discretization and method of lines gives the follow ordinary differential equation for each grid point

$$\frac{\mathrm{d}s_{ij}}{\mathrm{d}t} = \frac{D}{\Delta x^2} \left[ -4s_{ij} + s_{i-1,j} + s_{i+1,j} + s_{i,j-1} + s_{i,j+1} \right] + Rs_{ij} \left( 1 - s_{ij} \right)$$

Which we can express as the following nonlinear problem...

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





## **Numerical Solution**

- One nonlinear equation for each grid point
  - together they form a system of N=nx\*ny equations
  - Solve with Newton's method
- Each iteration of Newton's method has to solve a linear system
  - Solve with matrix-free Conjugate Gradient solver
- Solve the nonlinear system at each time step
  - This requires in the order of between 5-10 conjugate gradient iterations





- Don't worry if you don't understand it all!
- We don't need a deep understanding of the mathematics or domain problem to optimize the code
  - I often work on codes with little domain knowledge
- The mini-app has a handful of kernels that can be parallelized
  - And care was taken when designing it to make parallelization as easy as possible
- So let's look a little closer at each part of the code





## The Code

- There are two versions of the code
  - **–** C++
  - Fortran90
- Both codes have the same structure
  - Pick whichever version you are most comfortable with
- The code could be faster
  - Fortran version is faster than the C++ version
  - We avoided aggressive optimization to make it as easy as possible to understand.
  - Neither are they fine examples of design, which would get in the way of understanding.
- Tested with both Cray, GNU, Intel and PGI compilers





# Code Walkthrough

- There are three modules of interest
  - main.f90/main.cpp: initialization and main time stepping loop
  - linalg.f90/linalg.cpp: the BLAS level 1 (vector-vector)
     kernels and conjugate gradient solver
  - operators.f90/operators.cpp : the stencil operator for the finite volume discretization

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





# Linear algebra: linalg.f90/cpp

- This file defines simple kernels for operating on 1D vectors, including
  - dot product : x•y : ss\_dot
  - linear combination : z=alpha\*x + beta\*y : ss\_lcomb
- The kernels of interest start with ss\_xxxxx
  - ss == summer school
- For each parallelization approach that we will see (OpenMP, MPI, CUDA, ... etc), each of these kernels will have to be considered.





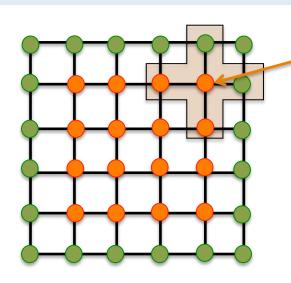
# Stencil operator: operator.f90/cpp

This file has a function/subroutine that defines the stencil operator

```
for j=2:ydim-1 for i=2:xdim-1 for i=2:xdim-1 f_{ij}=\left[-\left(4+\alpha\right)s_{ij}+s_{i-1,j}+s_{i+1,j}+s_{i,j-1}+s_{i,j+1}+\beta s_{ij}\left(1-s_{ij}\right)\right]^{k+1}+\alpha s_{ij}^{k}=0 end end
```



## Stencil: Interior Grid Points



interior points have all
 neighbours available

interior points

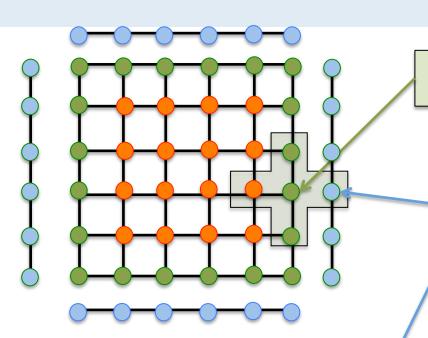
for j=2:ydim-1  $for i=2:xdim-1 \\ f_{ij} = \left[-\left(4+\alpha\right)s_{ij} + s_{i-1,j} + s_{i+1,j} + s_{i,j-1} + s_{i,j+1} + \beta s_{ij} \left(1-s_{ij}\right)\right]^{k+1} + \alpha s_{ij}^k = 0$  end



end



# Stencil: Boundary Grid Points



boundary points are missing 1 or 2 nieghbours

create 4 halo buffers, that
 hold "ghost" buffers
 bndN, bndE, bndS, bndW

east boundary

$$i=xdim$$

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

end





## Testing the code

Get the code, by checking it out from github

Compile and run

```
> make
> srun -n1 ./main 128 128 100 0.01 yes
```





# Testing continued...

- Compile
  - > make
- Run interactively (use salloc beforehand)
  - > srun ./main 128 128 100 0.01
  - the grid is 128 x 128 grid points
- take 100 time steps
- run simulation for t=0.01
- Or run batch job
  - > sbatch job.daint
  - ... when job is finished ...
  - > cat job.out

It is possible to choose parameters that will make the simulation fail to converge! The code should tell you gracefully that it was unable to converge.

increasing the spatial resolution may require increasing the number of time steps

#### Exercise

- Compile using the cray programming environment.
  - get time to solution for, note time to solution and total conjugate iterations in each case
    - 128 128 100 0.01
    - 256 256 200 0.01
    - 1024 1024 ??? 0.01
- Recompile using the GNU programming environment
  - this will require make clean to remove previous build:
  - > module swap PrgEnv-cray PrgEnv-gnu
  - > make clean
  - > make
    - rerun tests above and compare time to solution and the number of conjugate iterations for each case





## Output

The number of conjugate gradient iterations, which should always be constant for a given mesh size and time parameters. Can be used to check that changes to the code are still version getting the correct result. There will be small variations mesh due to the imprecise nature of floating point operations. time step 1 required 4 iterations for residual 7.21951e-07 required 4 iterations for residual 7.9975e-07 step step 99 required 12 iterations for res time to solution ste 100 required 12 iterations for simulation took 1.58408 seconds 8127 conjugate gradient iterations, at rate of 5130.43 iters/second 920 newton iterations Goodbye!



best way to compare different implementations



#### Visualize the answer

- The application generates two data files with the final solution: output.bin and output.bov
- There is a Python script that will show a contour plot of the solution

```
> ls output.*
output.bin output.bov
> module load python/2.7.6
```

> python plotting.py

> display output.png

requires X-windowing make sure you connect with "ssh -X"





## Exercise

- Visualize the output from the previous exercise
  - now is a good time to see if X-windows is working!
- If that worked properly, try visualizing output from different final times
  - srun –n1 ./main 128 128 100 0.0025
  - srun -n1 ./main 128 128 100 0.005
  - srun –n1 ./main 128 128 100 0.01





## Questions?

 You will become more familiar with the mini-app code over the summer-school.

