

# Parallel Programming Workshop

**Brought to you by**

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1 of 46*



# Registration

- **Please make sure you're signed in.**
- **Won't need a computer this morning**
  - *unless you need a calculator to add integers*



# Important Concepts

- Decomposition
- Scaling
- Speedup

We will jointly “discover” the meaning of these terms through experiment and group exercises – ease into programming only when necessary.



# Distributed Memory Programming

- Two main models for doing parallel programming:
- Distributed Memory – workers must talk with one another to get data.
- Shared Memory – Workers view the same memory space.

Each has different issues.

Take on Distributed Memory first.



# The Data Set

- Any confusion over the terms “integer” and “real” numbers?
- The data at hand consists of:
  - 50 data cards.
  - 5 integer numbers per card.
  - An integer card identifier.

**Set: 14**

**164**

**5**

**76**

**144**

**105**



# Exercise 1

- Desired analysis: summation over 4 cards
- Divide into groups.
- Each group needs a time keeper.

**Pay attention to the process!**



## Exercise 1 Outcomes

- What was the basic “unit of work” or task?
- What discreet steps were involved?

***Yea verily, computers are lowly beasts  
and must be instructed tediously.***



## Exercise 1 Summary

- Process had 3 distinct steps:
  - Hand out cards
  - Sum the numbers
  - Report results
- More formally:
  - Distribute work (tasks).
  - Perform work
  - Gather results





## Exercise 2 – Two Workers

- Repeat Ex 1, only with 2 people adding numbers.
- What changes?



# Added Workers

- What happened with more workers?
- The process changes a little:
  - Distribute work
    - How to do that? **Communicate!**
  - Perform work
  - Gather results
    - Gather partial results. **Communicate!**
    - Compute final result
    - Report result



## Exercise 3

- What happens with 3 workers?
- What happens with 4 workers?
- Could we use more than 4 workers?



## Exercise 3 Outcomes

- More workers => More communication
- Balanced work assignments?
- Task starvation? (run out of cards)
- How do the input and output compare with Ex 1?

*Everybody's talking at me, I don't hear a word their say'ng ...\**

*\* Fred Neil*



# Comment on Scaling

- How does parallel work speed up, i.e. “scale”?

$$S_p = \frac{T_1}{T_n}$$

$$S_{serial} = \frac{T_{serial}}{T_n}$$

- How efficient is it? Again, two types:

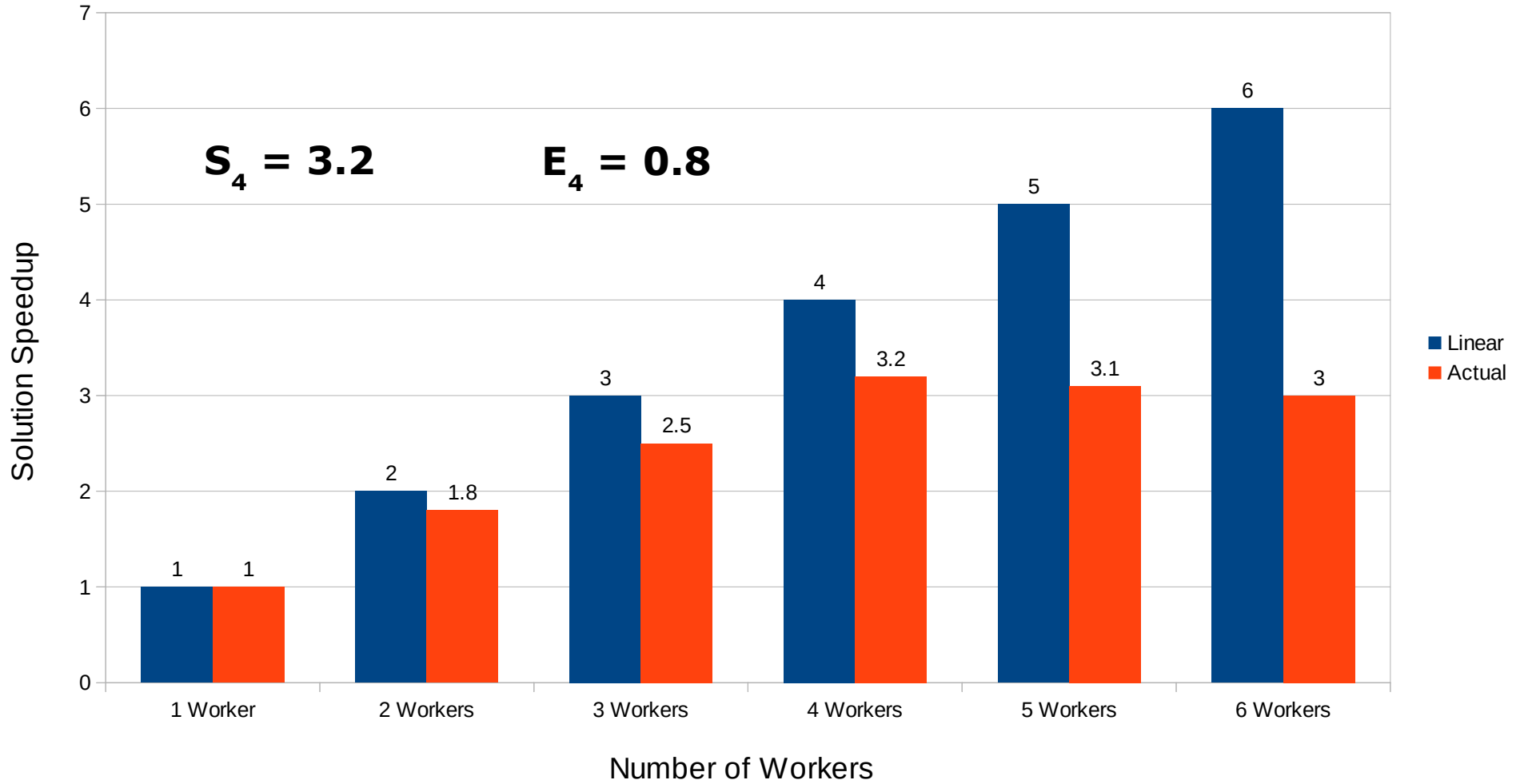
$$E_p = \frac{T_1}{nT_n}$$

$$E_{serial} = \frac{T_{serial}}{nT_n}$$

***Beware of “Lies, damn lies, and statistics . . .”***



Hypothetical Speedup Chart



## Overhead Expense

- 80% efficiency => 20% overhead.
  - If one hour on 5 computers, then 1 computer worth of power is unused!
- Constant trade-off between time-to-answer and expense, even if the usage seems "free".
- Time on most HPC systems is charged in core-hours (or *service units*), so low efficiency still costs more as service units are used up faster.



# Distributing Work (Data)

- Shared data?
  - Each worker has a copy
  - Each worker has an ID
  - Use ID to *compute* what to work on.
- Distributed data?
  - Head worker has all the data.
  - Head worker knows # of workers.
  - Head worker computes decomposition.
  - Head worker *sends pieces* to workers.





# Sharing Data

- Parallel file system – all workers see same data files.
- Broadcast – head worker broadcasts all data to all workers.



# Considerations

- How much time is required to communicate?
- Does machines have access to shared file systems?



## Concept Summary

When you approach problem to programming, ask yourself:

- What algorithm is required?
- How best to decompose the work?
- How is it suppose to scale?
- Minimize comm to get speedup.
- Test to see what has been achieved.



# Shared Memory Programming

- Distributed Memory Programming recap:
  - Each worker was isolated.
  - Sent or computed work decomposition info.
  - Sent data or shared via file system.
- What changes with Shared Memory Programming?
  - Workers part of same system (i.e. cores).
  - Each worker can see all data in memory.
  - *Communication* replaced by *coordination* of read/write access.



# Exercise 4

**Assume all workers can see all the data - how does summation task change?**

	<b>A</b>	<b>B</b>	<b>C</b>	<b>D</b>	<b>E</b>	<b>Sums</b>	
<b>1</b>	<b>6</b>	<b>3</b>	<b>13</b>	<b>78</b>	<b>35</b>		
<b>2</b>	<b>49</b>	<b>60</b>	<b>138</b>	<b>34</b>	<b>79</b>		
<b>3</b>	<b>59</b>	<b>108</b>	<b>108</b>	<b>188</b>	<b>110</b>		
<b>4</b>	<b>137</b>	<b>50</b>	<b>4</b>	<b>167</b>	<b>189</b>		
<b>5</b>	<b>83</b>	<b>136</b>	<b>215</b>	<b>26</b>	<b>140</b>		
<b>6</b>	<b>0</b>	<b>187</b>	<b>77</b>	<b>216</b>	<b>51</b>		
							<b>Total</b>



## Exercise 4 Outcomes

- Benefits?
- Difficulties?



# Concept Summary

- Shared memory lets all processors see all data, it is just there – no work to distribute it. BUT, need to coordinate changes!
- Shared Memory Model is growing in popularity as more cores per node become available, and new devices such as GPUs become common place – multi-core PCs use shared memory.
- Hybrid or Heterogeneous models are becoming important as the needed to combine Shared and Distributed models increase.



# Parallel Thinking

- What kind of questions do you need to consider when approaching a new program?
- Algorithm – numerical stability? programmability?
- Data size – memory needs?
- Machine architecture – shared/distributed/both?
- Code lifetime – save FTE's or machine hours?
- Choice of language?
- Choice of tools?





**Break**



# The Laplace Heat Equation

- For a “real” problem, consider how to go about solving the Laplace Heat Equation in 2-D. Idea is to determine the temperature at any point on a surface, given the temperature at the boundaries:



# Formal Solution

The solution must satisfy:

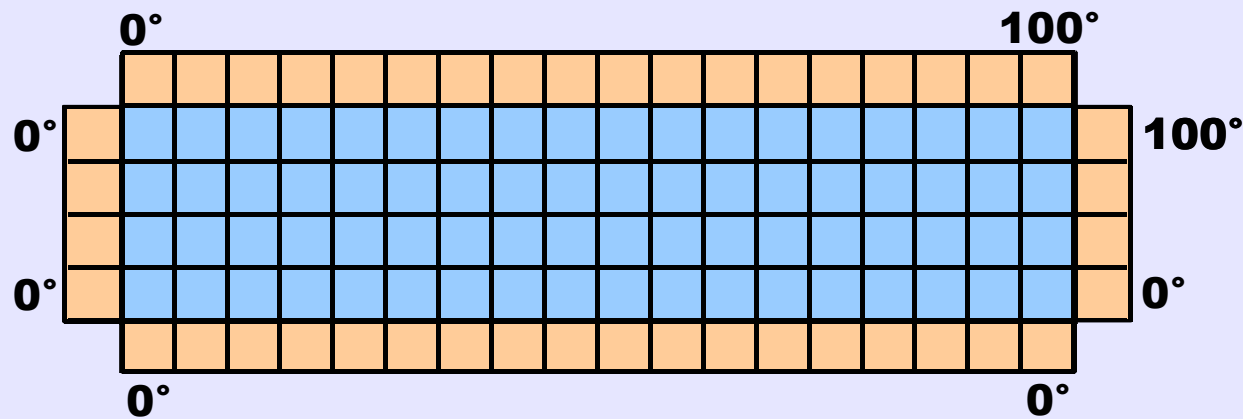
$$\nabla^2 \varphi = 0$$

with the application of Dirichlet boundary conditions  
(constant values around edge of region.)



# The Serial Solution

Subdivide the surface into a mesh of points, add boundary points.

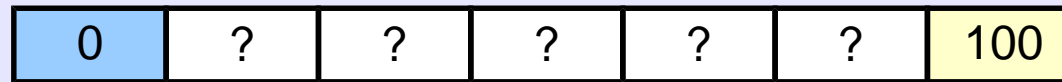


Apply the following *5-point stencil* iteratively until the temperature stops changing (new temp approximates old temp) to interior only:

$$T_{i,j}^{n+1} = 0.25 * (T_{i-1,j}^n + T_{i+1,j}^n + T_{i,j-1}^n + T_{i,j+1}^n)$$



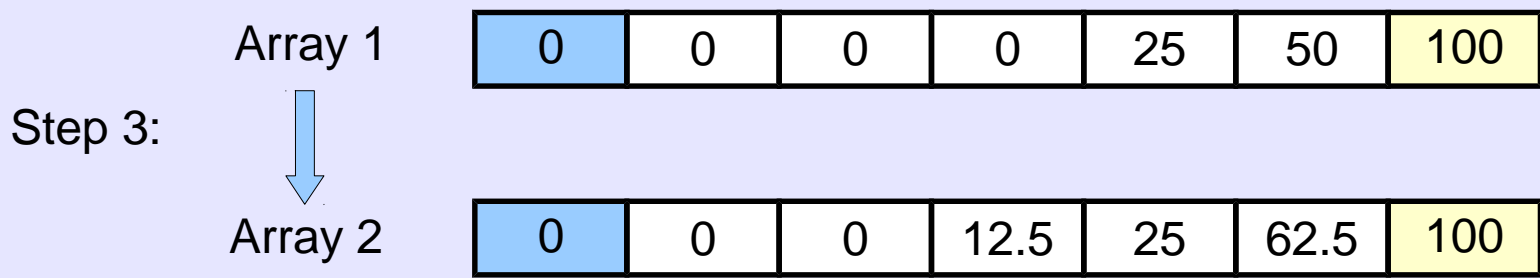
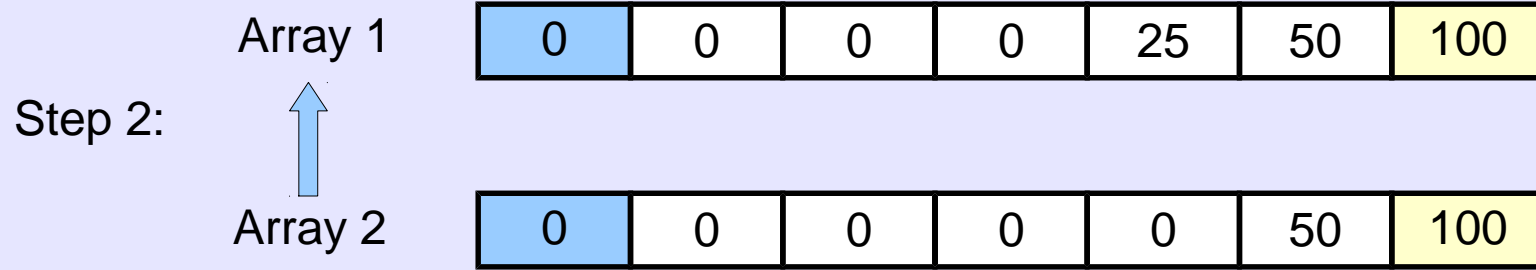
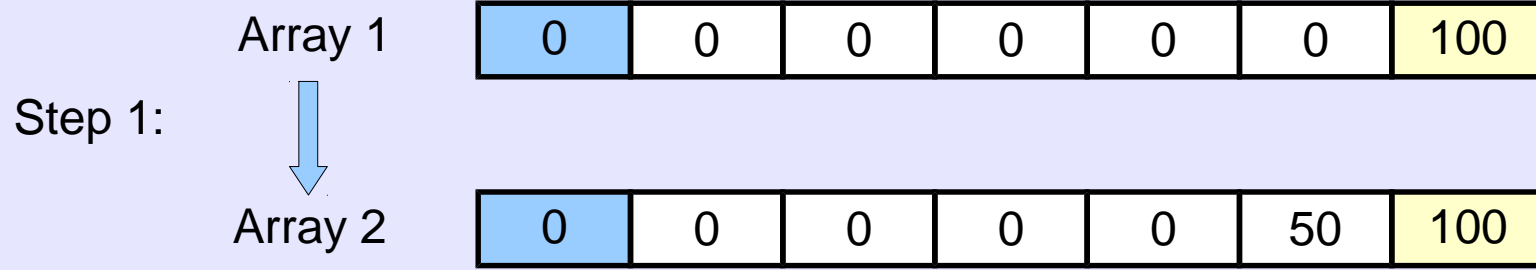
## Exercise 5: 1-D Problem



$$T_i^{n+1} = 0.5 * (T_{i-1}^n + T_{i+1}^n)$$

Discuss programming this problem in your group.





## Exercise 5: Solution

70 iterations to reach 0.001% convergence bound.

0	16.6661	33.3324	49.9988	66.6658	83.3327	100
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## Exercise 6



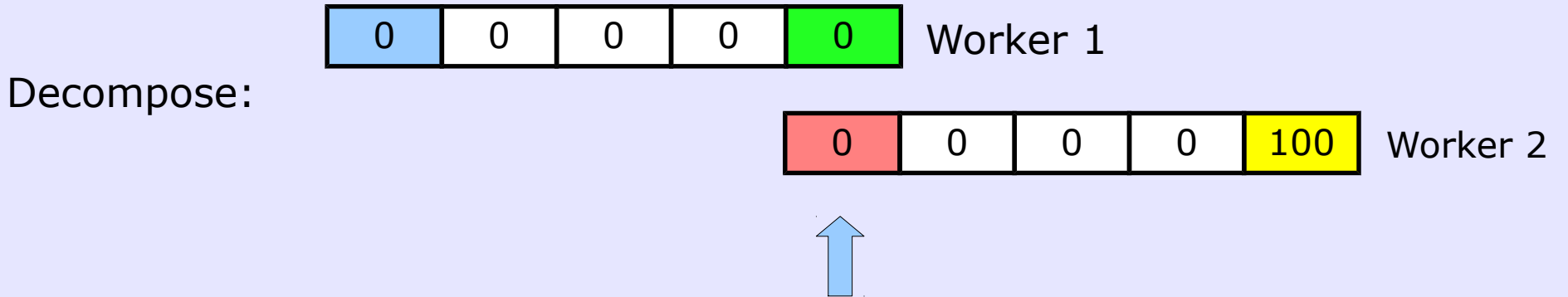
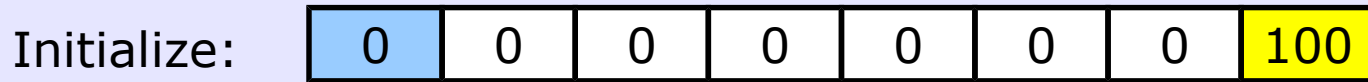
**Now the question is, how would we do this in parallel?**

**Need one small modification, so try using 2 workers first.**





# Process Start

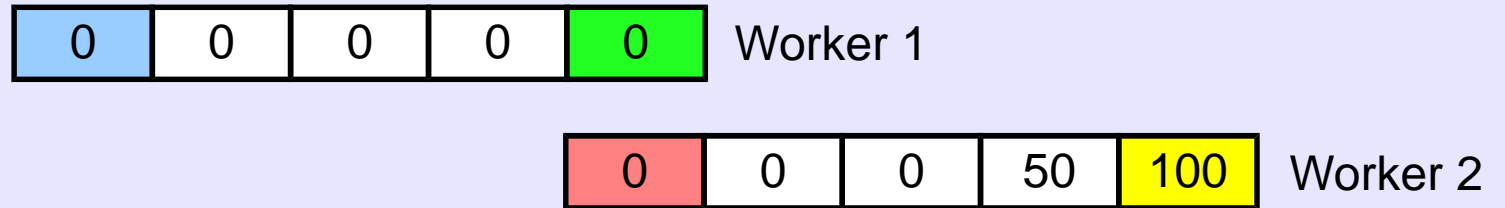


"Ghost" or overlapped cells.

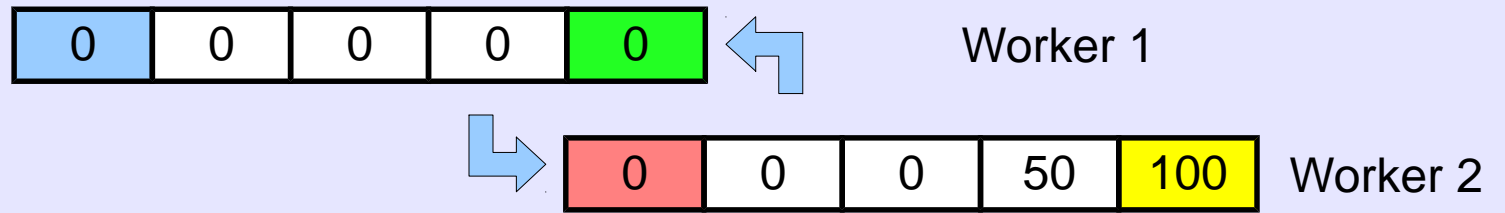


# Process Iteration

Compute:



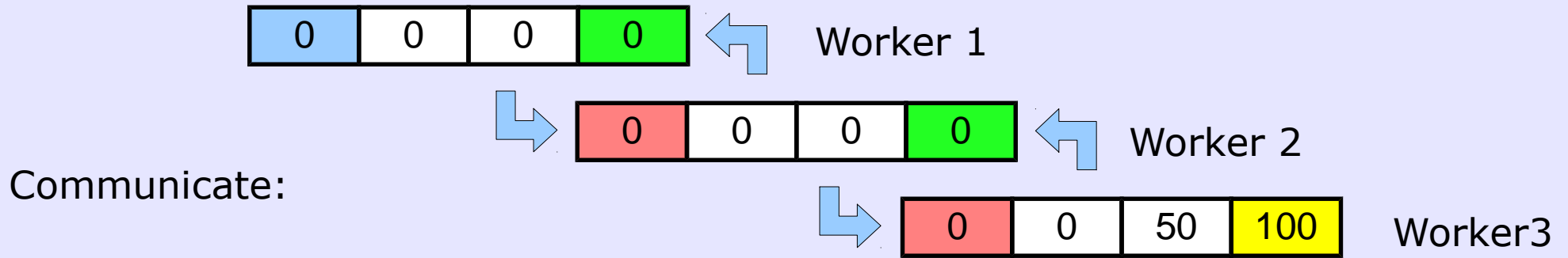
Communicate:



Lather, Rinse, Repeat.



# What Would 3 Workers Involve?



Workers in the middle have to communicate intermediate results to neighbors on both sides!

**Number of workers limited by problem size!**



# Serial Program

- Grab a copy of the program named:  
`/work/jalupo/laplace_solver_serial.f90`
- Open with "less" or "vi" so you can follow along.
- Anyone have trouble reading Fortran?
- Anyone not know how to compile and run a Fortran program?



## Main Components

- `program laplace_main` – program main line.
- `subroutine laplace` – the actual solver. It also allocates memory to hold the 2-D mesh based on the requested rows and columns.
- `subroutine initialize` – sets the internal temperatures to 0.
- `subroutine set_bcs` – sets up the boundary conditions.



## Compiling Fortran

- Here is a quick summary of how to compile and run this particular program (assumes default environment):

```
$ ifort -o laplace laplace_solver_serial.f90
```

```
$ ./laplace
```

- You should see the following line of text on your screen:

```
Usage: laplace nrows ncols niter iprint relerr
```

Now try executing the program with some real numbers:

```
$ ./laplace 100 200 3000 300 0.001
```



## Results of Run

```
$ ./laplace 100 200 10000 3000 0.01
```

Solution has converged.

Iterations: 2241

Max error: 0.01

Total time: 0.079s

What if the problem gets bigger, and error condition was changed to 0.001?



# Higher Accuracy Run

```
$ ./laplace 1000 1000 30000 1000 0.001
```

Solution has converged.

Iterations:	29812
Max error:	0.001
Total time:	60.546s





# Why go to parallel?

**What if this was only part of a simulation and the temperatures changed 25,000 times?**

**Even though 1 solution taking 1 second seems fast, 25,000 solutions would take 7 hours!**

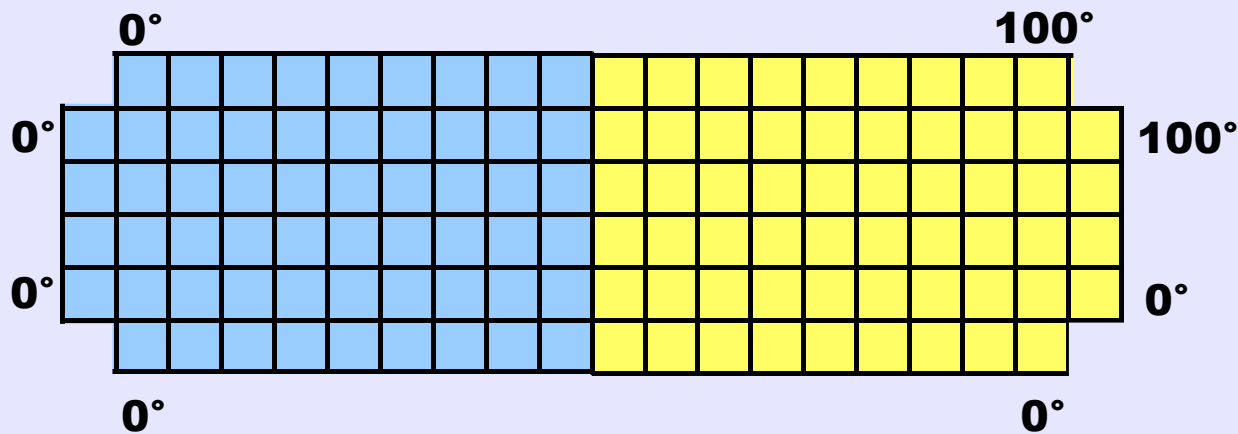
**Can it be done in parallel to speed up the over all simulation time?**

**How do we approach the solution in parallel?**



# Decomposition

Assuming 2 processors, let's divide the surface in half.



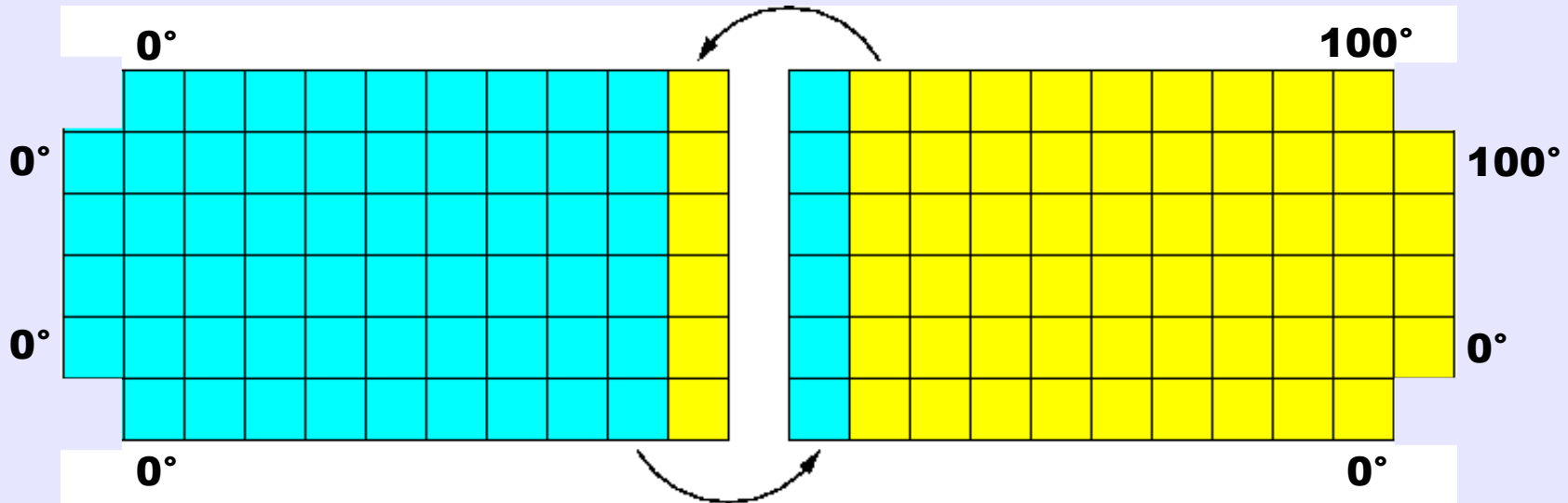
What overhead do we have to consider adding to make this give the same answer?



# Ghost Cells

Process 1

Process 2



# Overhead

- Breaking up the problem so multiple processes can work on it introduces *overhead*:
  - Logic must be added so each process knows which part of the mesh it is expected to work on. This directly impacts how the code will start up.
  - Communication must be added so data from adjoining regions can be properly updated.
  - Code must be added so the final results can be communicated. This directly impacts how the code will report results and terminate.
- A serial program is not the same as a parallel program running on 1 processor!



# Compute/Communication Bound

- Clearly, if you increase the number of processes working on this problem, the amount of communication required increases.
- With a few processes, this problem exhibits the property of being *compute bound*.
- When the number of processes approach the number of mesh points, it becomes *communication bound*.
- All parallel programs exhibit one form or the other depending on the problem specifics.



# LUNCH

