

Communication Optimizations

Didem Unat COMP 429/529 Parallel Programming

Communication Optimizations

- Reduce the number of messages or total message size
 - Reduces communication cost
- Hide communication
 - Hide the communication overhead behind useful computation
- Avoid communication
 - Doesn't perform communication at the expense of more computation
- Combine (aggregate messages)
 - Instead of sending many small messages, send few large messages
- These methods can be combined

Reducing Communication Cost

- Reformulate MPI applications to reduce communication cost
 - Major restructuring of a program
 - Code can become very messy
 - Split phase coding with non-blocking MPI calls

Split-Phased Execution

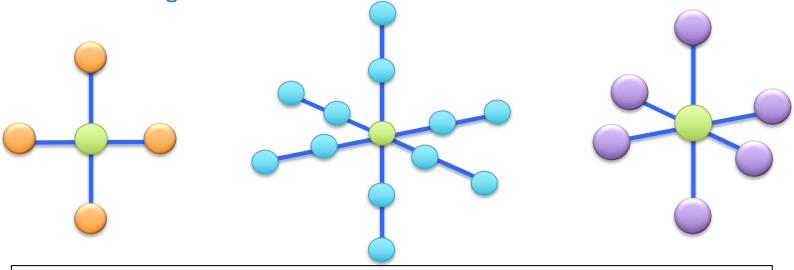
- Phase 1: initiate communication with Irecv(), Isend()
- Phase 2: synchronize Wait()
- Perform unrelated computations between the two phases
- Each pending irecv()s and isend()s must have a distinct buffer

OVERLAP	NO OVERLAP
Irecv(x,req) ISend() Compute(y) Wait(req) Compute(x)	Irecv(x) ISend() Wait(x) Compute(x) Compute(y)

Stencil Shape

- Stencil shape can be different depending on the numerical algorithm used
- Here 5-point in 2D, 13-point and 7-point in 3D are shown

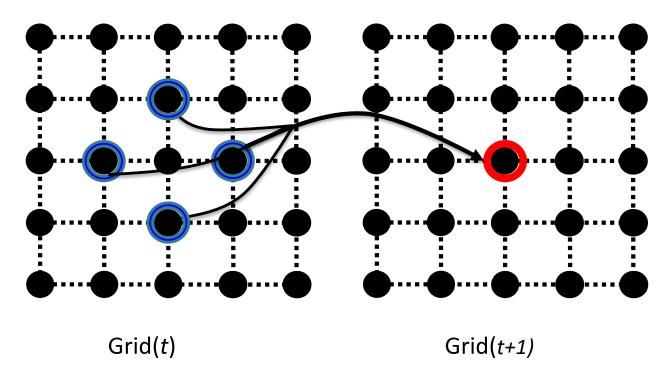
 Next value of green point depends on the current value in the nearest neighbors



```
//image smoothing example using 5-point stencil
for iter = 1 : nSmooth
  for (i,j) in (0:N-1, 0:N-1)
        Imgnew [i,j] = (Img[i-1,j]+Img[i+1,j]+Img[i,j-1]+Img[i, j+1])/4
  Swap(Imgnew,Img)
```

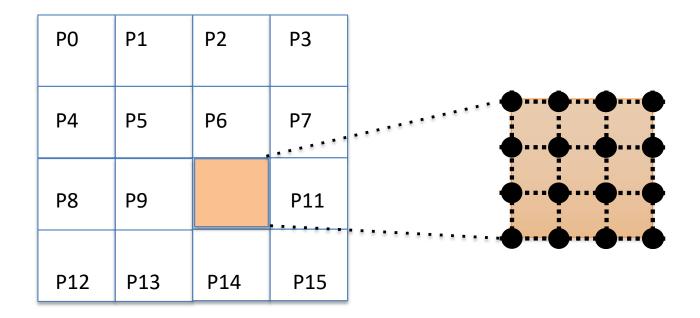
Cell Update

- Since we are updating the next value of a cell from its neighbors, typically we need to keep two grids around
 - Current and Next Grid
- There are techniques to implement with one grid, they are called update-in-place methods



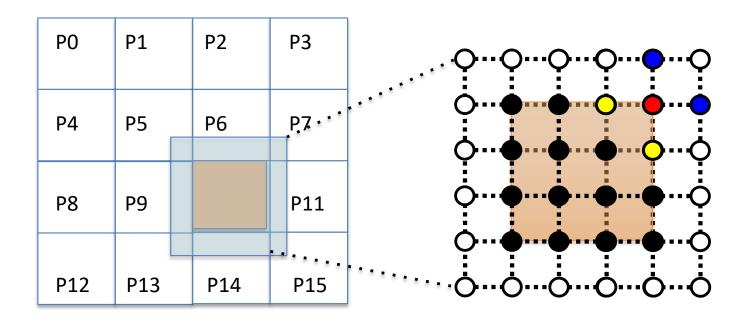
Data Dependences

- Each process has its own data partition
- Do we need any data points from neighboring processes?



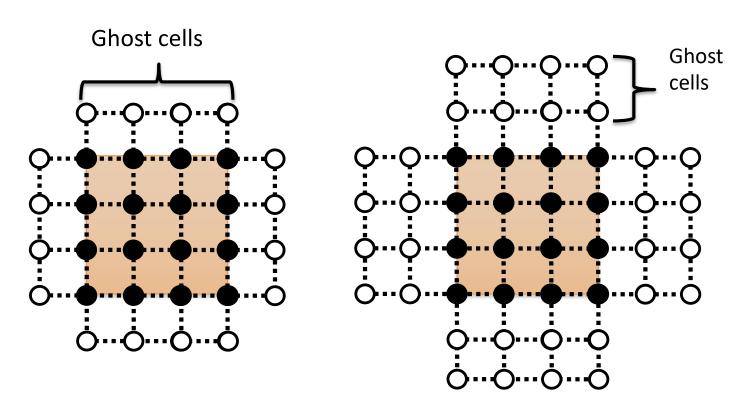
Data Dependences

- In order to update red, we need yellow points (local) and blue points from other processes (remote)
 - Too expensive to communication individual array values one at a time
 - "Ghost" cells hold a copies off-process values



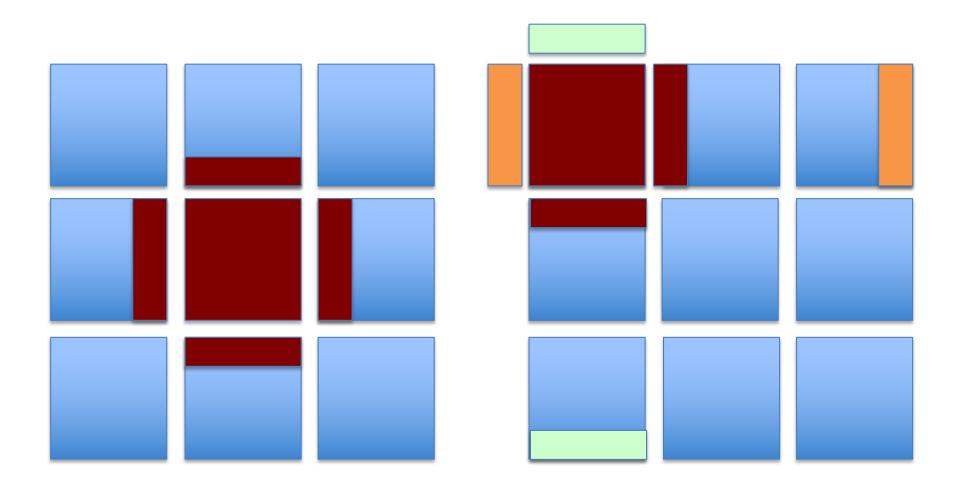
Ghost Cells (Halos)

- Ghost cells are allocated for the information needed from other processes
 - Thus each processor allocates (N+ 2*ghosts) d space to accommodate ghost region
 - Depth of ghost region depends on the numerical method used



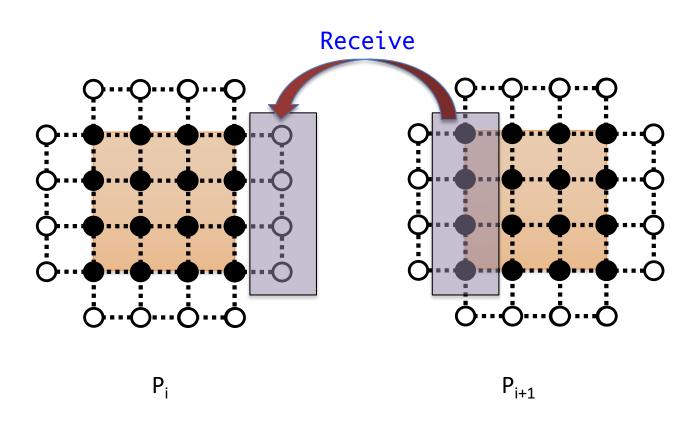
Boundary Updates

• An example of periodic boundary conditions:



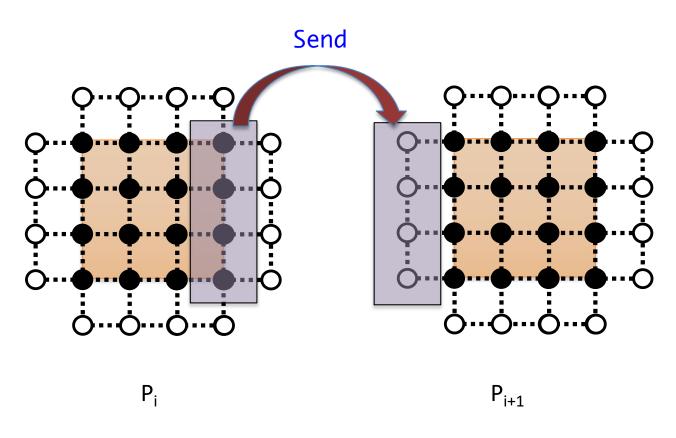
Ghost Cell Exchange- Receive

- Need to update ghosts every iteration so that a process has the latest values
- This is done with message exchange between processes



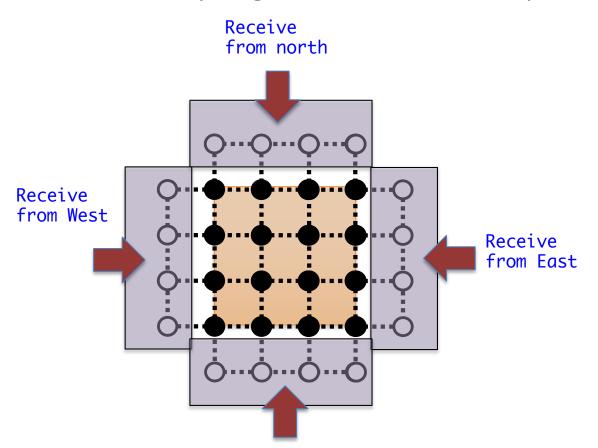
Ghost Cell Exchange- Send

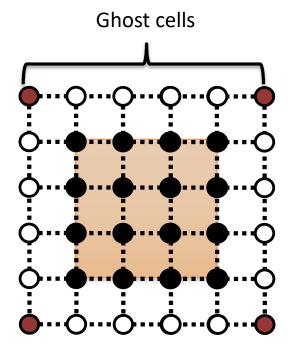
- Need to update ghosts every iteration so that a process has the latest values
- This is done with message exchange between processes
 - Note that the elements may not be contiguous in memory need to pack a message at the source and unpack at the destination



Msg Exchange with All Neighbors

- How many processors involved in communication depends on the numerical algorithm and domain decomposition
 - In 2D, typically sends/receives involve 4 neighbors (N, S, W, E)
- How many neighbors are there for 3D problems?





If the computation uses the corner cells in the update, then need to exchange ghosts with 4 more processes: NE, NW, SE and SW

Communication Cost

- Communication performance can be a major factor in determining application performance
- Message passing time = $\alpha + N/\beta$

```
\alpha = message startup time and overhead (latency)
```

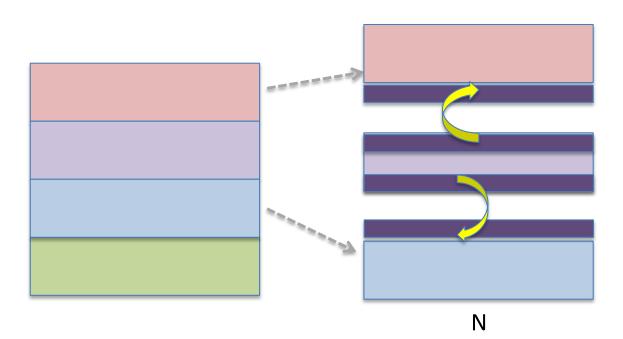
 β = network bandwidth (bytes/sec)

N = message size

- Short messages: startup term dominates
- Long messages: bandwidth term dominates

Communication Cost for 1D Geometries

- Assumptions P processors divides input size N evenly
 - 1 word = double precision floating point = 8 bytes
- For horizontal strips, data are contiguous
 - Row-major C arrays
- Message Size = 16N bytes (sent to North + sent to South) per processor $Time(comm) = 2(\alpha + 8N/\beta) \text{ per processor}$



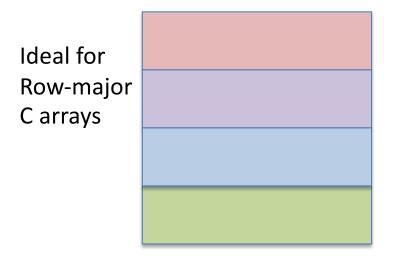
α :latency

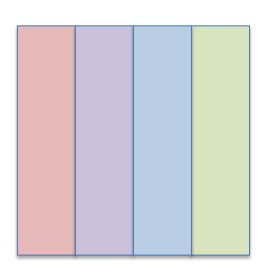
β: bandwidth

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Communication Cost for 1D Geometries

- For horizontal strips, data are contiguous
 - Row-major C arrays
- For vertical strips, data are not contiguous in memory
- This model doesn't take into account the message packing overhead



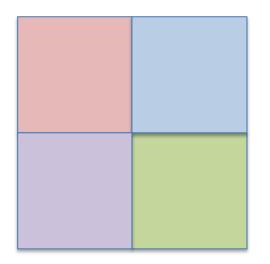


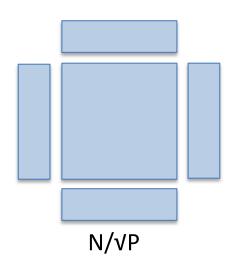
Ideal for Column-major arrays

Communication Cost for 2D Geometries

- Assumptions
 - VP divides N evenly
 - ▶ 1 word = double precision floating pt. = 8 bytes
- Ignore the cost of packing/unpacking message buffers
- Message size sent in bytes for processor is 4*8N/√P

Time(comm) =
$$4(\alpha + 8N/(\beta\sqrt{P}))$$





Process Geometry

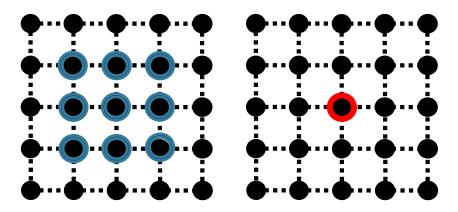
- Block decomposition is harder to implement than strip decomposition
- Any good reasons why an application should support block decomposition?
- Typically strip decomposition outperforms block decomposition resulting in lower communication times when

$$2(\alpha+8N/\beta) < 4(\alpha+8N/(\beta\sqrt{P}))$$

• Optimal process geometry depends on the N, β and α .

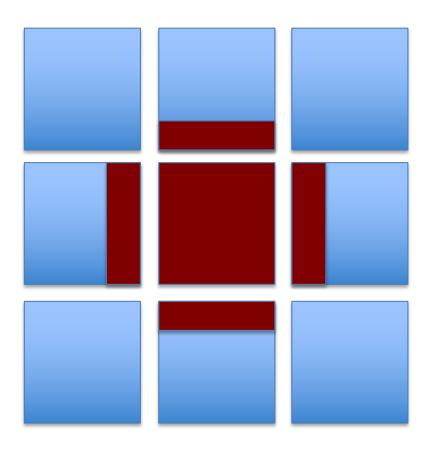
Question

```
for (i=1:N)
for (j=1:N)
Unew[i,j] = (-20*U[i,j] + 4*[U[i,j+1]+U[i,j-1] + U[i+1,j]+U[i-1,j]) +
U[i+1,j+1] + U[i+1,j-1] + U[i-1,j+1] + U[i-1,j-1]) / (6*h)
swap(U, Unew)
```

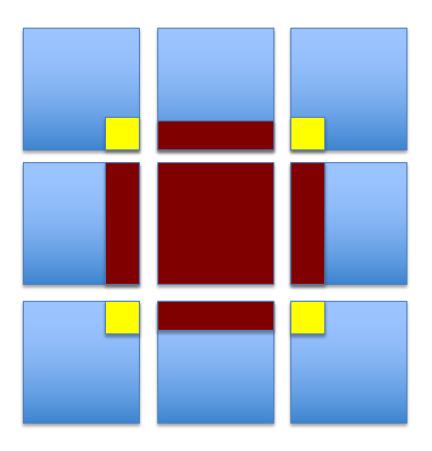


- 9-point stencil
- Case A: Cost model for ghost cell exchange with 8 neighbors
- Case B: Cost model for ghost cell exchange with 4 neighbors

5-point Stencil Case

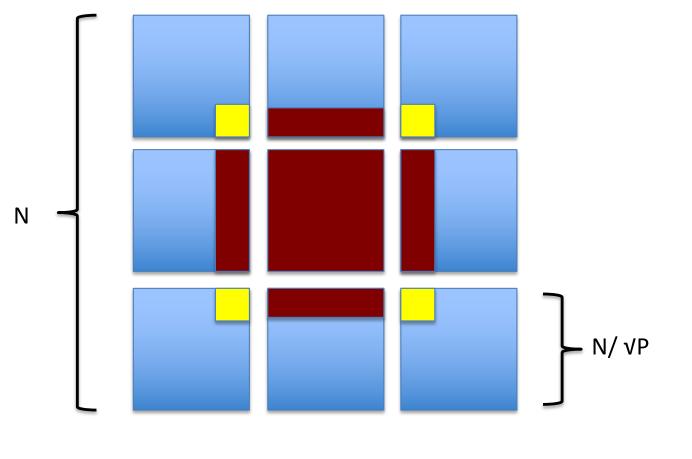


Time(comm) =
$$4(\alpha+8(N/\sqrt{P})/\beta)$$

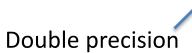


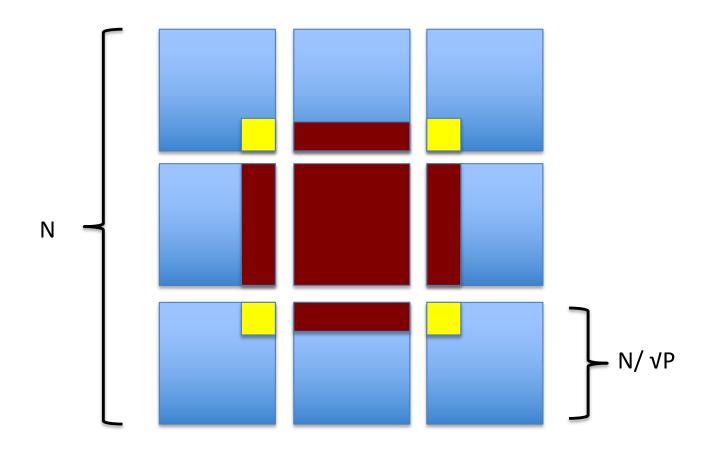
Need to get data from 8 different neighbors

Input size N*N, number of processors is P

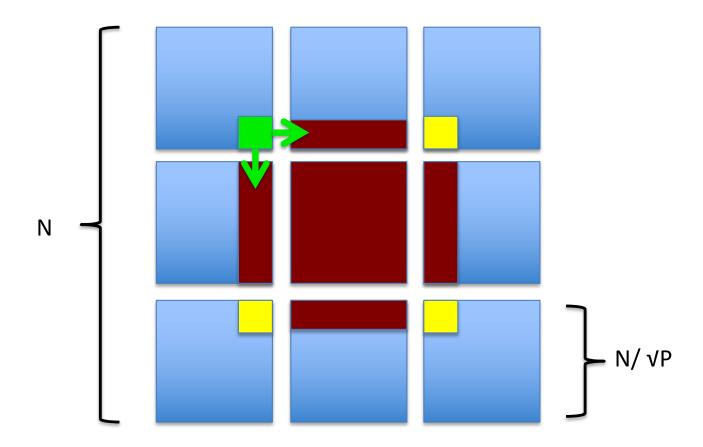


Comm Cost=
$$4 (\alpha + 8 (N/VP)/\beta) + 4(\alpha + 8 * 1/\beta)$$

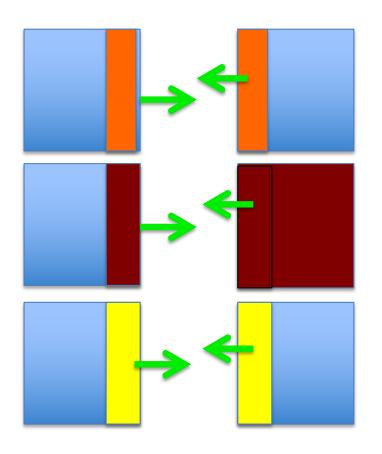




Can we do better? Exchange message with only 4 neighbors!



Note that the green point is sent to two of my neighbors! Why don't I get that point from my North or West neighbor rather than getting it from NW neighbor?



Showing only exchange with W neighbors.

Assume we are also exchanging with E neighbors in this step

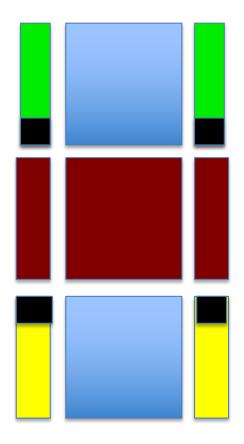
Do message exchange **in two steps**. First everyone sends ghost cells to W and E or (to N and S)

Cost of Step 1

Comm Cost= 2 (
$$\alpha$$
 +8 (N/ \forall P)/ β)

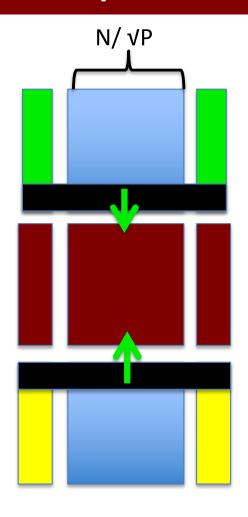
2 because we only communicated with E and W

End of First Step, we have new ghost cells on our W and E sides. These include the corner cells that we needed!



In the second step, exchange with N and S but slightly larger data (N/VP + 2) that include the corner points

End of First Step, we new ghost cells on our W and E sides. These include the corner cells that we needed!



Note that step 1 has to finish before we start step 2

In the second step, exchange with N and S but slightly larger data (N/VP + 2) that include the corner points

Step 2

Comm Cost for Step 2= 2 (
$$\alpha$$
 +8 * (N / \sqrt{VP} + 2) * 1/ β)

Two more elements

Then Combined Step 1 and 2

Comm Cost =
$$4 (\alpha + 8 (N/VP)/\beta) + 4 * 8 * 1/\beta$$

Comparison: Case A - Case B = 4α thus case B is better

Pseudocode

```
while (t < tFinal)</pre>
      Compute(...);
      Step1:Post communication for East&West neighbors
      Wait for completion of Step1
      Step2:Post communication for South&North neighbors
      Wait for completion of Step2
      t++;
```

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Note that there is no communication hiding here.

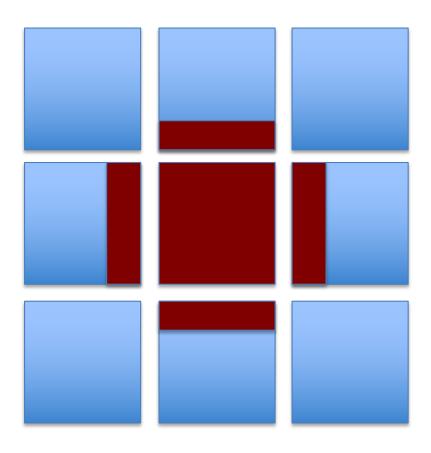
Hiding Communication Overhead

- Non-blocking communication calls can be used to hide communication overhead
 - Also called hiding communication latency
- When used correctly, these primitives are capable of overlapping communication overheads with useful computations.
- What is the maximum speedup you can get by hiding communication overhead?

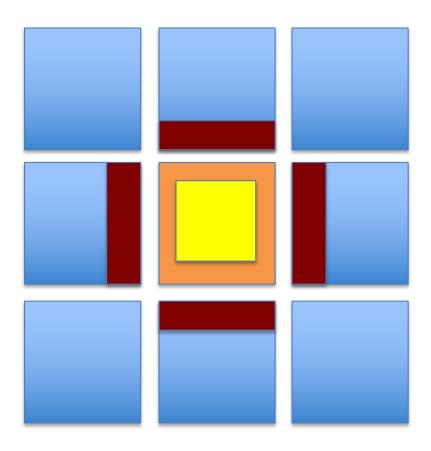
Nearest neighbor exchange in a ring topology with Non-Blocking Messages

```
#include "mpi.h"
#include <stdio.h>
main(int argc, char *argv[]) {
int numtasks, rank, next, prev, recvbuf[2], sendbuf[2], tag1=1, tag2=2;
MPI Request regs[4];
MPI Status stats[4];
                                                0 2 1 2 2 2 3 2 4 2 5 2 ... 2N
MPI Init(&argc,&argv);
MPI Comm size(MPI COMM WORLD, &numtasks);
MPI Comm rank(MPI COMM WORLD, &rank);
initBuffers(sendbuf);
prev = rank-1;
next = rank+1;
if (rank == 0) prev = numtasks - 1;
if (rank == (numtasks - 1)) next = 0:
MPI Irecv(&recvbuf[0], 1, MPI INT, prev, tag1, MPI COMM WORLD, &reqs[0]);
MPI Irecv(&recvbuf[1], 1, MPI INT, next, tag2, MPI COMM WORLD, &reqs[1]);
MPI_Isend(&sendbuf[0], 1, MPI_INT, prev, tag2, MPI_COMM_WORLD, &reqs[2]);
MPI Isend(&sendbuf[1], 1, MPI INT, next, tag1, MPI COMM WORLD, &reqs[3]);
{ do some work }
                                         Do some useful work to hide the communication
                                     overhead. Work cannot depend on the message buffers!
MPI Waitall(4, regs, stats);
MPI Finalize();
```

 How can we hide the communication in stencil computation?



- Separate the computation as inner (yellow region) vs outer (orange)
- Outer region requires ghost cells from neighbors but
- Inner region doesn't depend on other processors



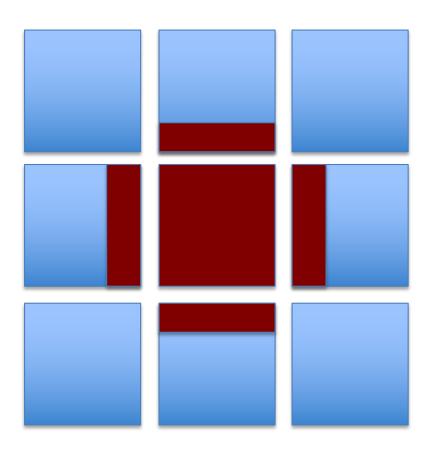
Pseudocode for Comm Hiding Version

```
while (t < tFinal)</pre>
      ComputeOrange(...);
      Post asynchronous comm with neighbors
      ComputeYellow(); // while waiting for comm
      Wait for completion of comm
      t++;
```

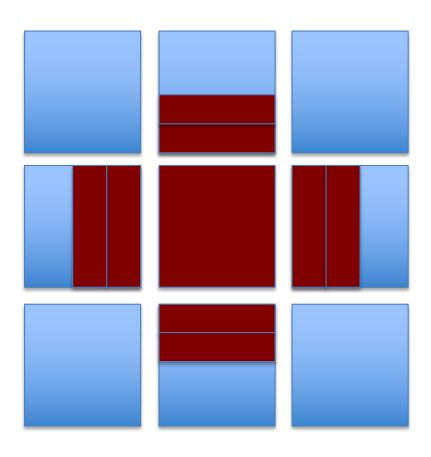
 We can even go more fine-grained and compute N, post comm for N, compute S, post comm for S,etc

- Communication avoiding version of stencil method can be implemented by having more ghost cells on each side
 - For example, instead of 1 ghost cell, keep a deeper cost region e.g. 4 ghost cells
 - Exchange message every 4 iterations instead of every
 - This reduces the number of messages but not the message size
 - The goal is to compute for a longer period of time before performing a communication

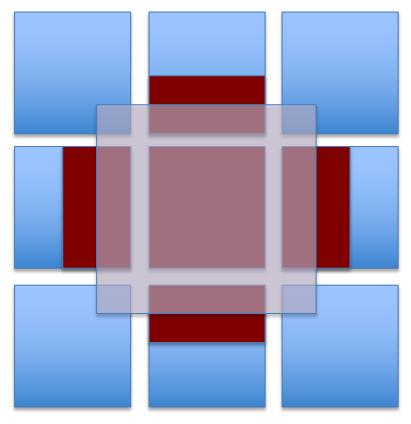
• Ghost cells =1



• Ghost cells =2

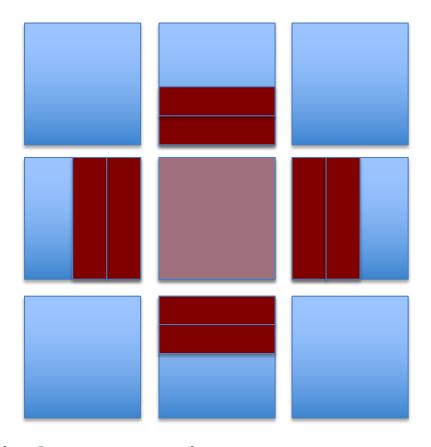


Compute slightly larger area using the outermost ghost cells



 Note that we redundantly compute the borders because those elements are also computed by another rank

 Compute slightly larger area using the outermost ghost cells and then shrink the area computed and use inner ghost cells



 This works! Why? Because data movement is a lot more expensive than computing

Acknowledgments

- These slides are inspired and partly adapted from
 - -Metin Aktulga (Michigan State Univ.)
 - -Scott Baden (UCSD)
 - —The course book (Pacheco)
 - -https://computing.llnl.gov/tutorials/mpi/