## High-Performance Computing Lab for CSE

Project 4 - Parallel Programming using the Message Passing Interface MPI Due date: 26 April 2021, midnight.

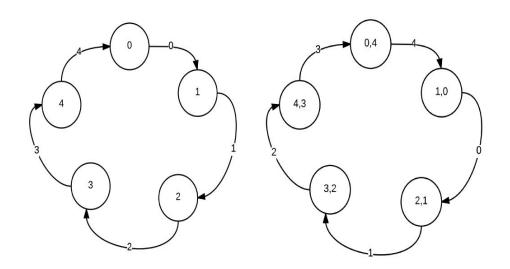
### Project 4: Parallel Programming using MPI

- 1. Ring addition using MPI (10 Points).
- 2. Ghost cells exchange (15 Points).
- 3. Mandelbrot set (20 Points).
- 4. Option A: Parallel matrix-vector multiplication and the power method (40 Points).
- 5. Option B: Parallel PageRank Algorithm and the power method (40 Points).

### Ring addition using MPI

• Each process:

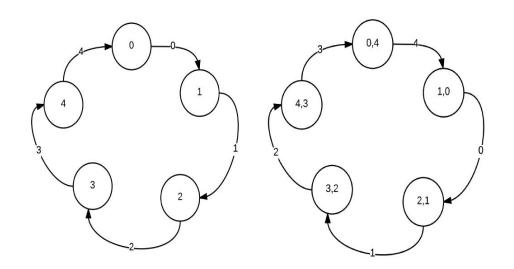
Receives from the left, sends to the right.



First two iterations of the ring sum algorithm.

### Ring addition using MPI

• After n iterations, each process has the sum of all ranks.



First two iterations of the ring sum algorithm.

### Ring addition using MPI

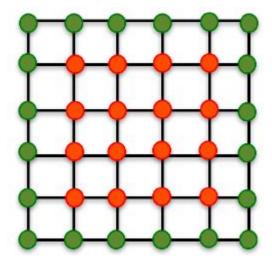
Solve the following tasks:

- Determine left/right neighbors of each process.
- Implement a ring addition code.

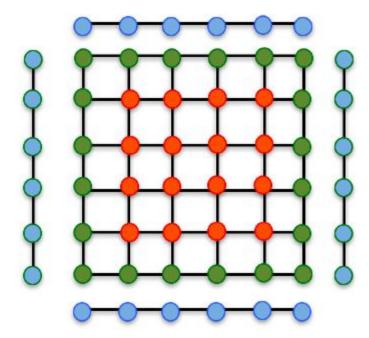
Result of the parallel computation with n = 4 should be:

```
Process 0: Sum = 6
Process 1: Sum = 6
Process 2: Sum = 6
Process 3: Sum = 6
```

- Each process has a 6x6 local domain.
- The borders must be sent to the neighbors (top, bottom, left, right).



• To handle the copies, we create ghost cells. Extended domain of size  $(6 + 2) \times (6 + 2)$ 



• We assume an  $n \times n$  grid of processes in a Cartesian topology.

0	1	2	3
4	5	6	7
8	9	10	11
12	13	14	15

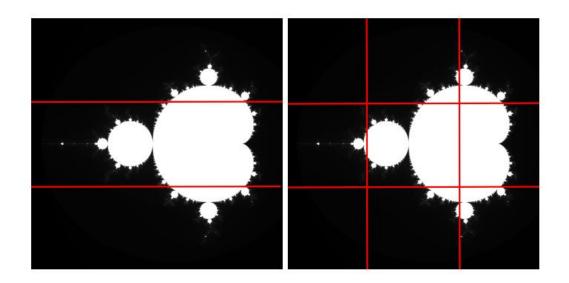
#### Solve the following tasks:

- Create a Cartesian 2-dimensional communicator with periodic boundaries.
- Create a derived data type for a column border.
- Exchange ghost cells with neighboring cells and verify correctness.
- Work with 16 processes.

### Mandelbrot Set

Create partitioning of the domain.

Each process computes its local portion of Mandelbrot set.



#### Mandelbrot Set

Solve the following tasks:

- Implement createPartition and updatePartition.
- Implement createDomain.
- Send the local domain to the master process.
   Compare the output with sequential program.
- Analyze the performance.
   Discuss about benefits and load balancing observed.

The power method is used to find the largest eigenvalue of a matrix.

```
[n,n] = size(A);
x = rand(n,1);
for i = 1:1000
    x = x / norm(x);
    x = A * x;
end;
lambda = norm(x);
```

Your task is to write a parallel C/MPI code that does the same computation.

#### What to implement

- generateMatrix
- powerMethod. Calls norm and matVec.
- main. Calls generateMatrix and powerMethod.

Your task is to write a parallel C/MPI code that does the same computation.

#### Where's the data

Assume that n, the number of rows and columns is divisible by p, the number of processors.

Each processor should generate its own rows of the matrix.

#### What experiments to do

#### Strong scaling analysis.

Fix a value for n. Run your code for p = 1, 4, 8, 12, 16, 32, 64.

Report running time and parallel efficiency.

#### Weak scaling analysis.

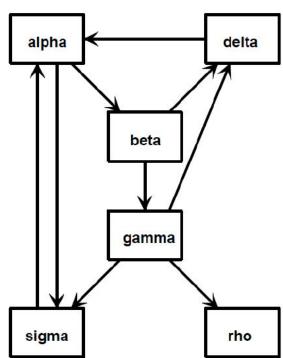
Choose n proportional to sqrt(p). In theory, work remains constant as you increase p.

Report running time and parallel efficiency.

PageRank Algorithm.

A random surfer goes through the websites.

What is the probability of going from one to the other?

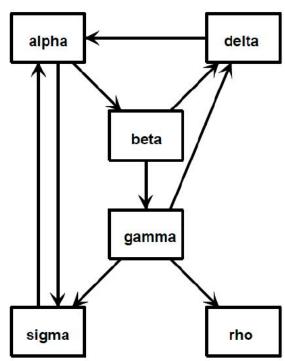


#### PageRank Algorithm.

Ranking websites according to the state vector of a Markov chain:

$$x = Ax$$

The largest eigenvalue of A is 1.



Sparse representation of matrix A

$$A = pGD + ez^T$$

Power method can be written as

```
G = p*G*D

z = ((1-p)*(c^=0) + (c==0))/n;

while termination_test

x = G*x + e*(z*x)

end
```

A serial implementation is provided.

Your task is to make the necessary MPI changes to obtain a parallel implementation.

Benchmark your code with the three Stanford Network Dataset Collection (SNAP) examples provided.

Different sparsity patterns. Comment on the results obtained.

Computation, synchronization and communication time.

