Parallel Computing I

Cluster: All-Gather, Scalability and Pipelining, and Overlapping

Predict the (past and/or future) motion of ${\it N}$ objects, given their motion at the present time.

- "motion" includes position as a function of time
- ▶ motion of one object influenced by external fields and other objects

Predict the (past and/or future) motion of ${\it N}$ objects, given their motion at the present time.

- "motion" includes position as a function of time
- motion of one object influenced by external fields and other objects

Gravitational N-body problems:

motion of astronomical bodies (planets, stars, star clusters, galaxies),
 due to mutual gravitational attraction

Electro-magnetic **N**-body problems:

motion of charged particles (protons, antiprotons),
 due to mutual attraction/repulsion and magnetic fields

Predict the (past and/or future) motion of ${\it N}$ objects, given their motion at the present time.

- "motion" includes position as a function of time
- motion of one object influenced by external fields and other objects

For systems of three or more objects, no analytic formula exists. Instead, must simulate the motion.

Predict the (past and/or future) motion of N objects, given their motion at the present time.

- "motion" includes position as a function of time
- motion of one object influenced by external fields and other objects

For systems of three or more objects, no analytic formula exists. Instead, must simulate the motion.

Simulations make simplifying assumptions:

- objects are idealized point particles (have mass and charge, but no volume)
- geometry assumed to be Euclidean
- motion and forces assumed to be Newtonian

Predict the (past and/or future) motion of ${\it N}$ objects, given their motion at the present time.

- "motion" includes position as a function of time
- ▶ motion of one object influenced by external fields and other objects

For systems of three or more objects, no analytic formula exists. Instead, must simulate the motion.

Simulations make simplifying assumptions:

- objects are idealized point particles (have mass and charge, but no volume)
- geometry assumed to be Euclidean
- motion and forces assumed to be Newtonian

Brief mathematics and physics review.

N-body Problems: Euclidean Geometry

Vectors:

 \blacktriangleright two-dimensional (2D) vectors, consist of (orthogonal) x and y components

$$\vec{a} = \langle a_x, a_y \rangle$$

Vector arithmetic:

vector addition

$$\vec{a} + \vec{b} = \langle a_x + b_x, a_y + b_y \rangle$$

vector subtraction

$$\vec{a} - \vec{b} = \langle a_x - b_x, a_y - b_y \rangle$$

scalar product

$$c \cdot \vec{a} = \langle c \cdot a_x, c \cdot a_y \rangle$$

vector magnitude

$$|\vec{a}| = \sqrt{a_x^2 + a_y^2}$$

N-body Problems: Newtonian Physics

Physical characteristics:

▶ mass of particle *i*

 m_i

position of particle i at time t

$$\vec{p}_i(t) = \langle p_{i,x}(t), p_{i,y}(t) \rangle$$

velocity of particle i at time t

$$\vec{\mathbf{v}}_i(t) = \langle \mathbf{v}_{i,x}(t), \mathbf{v}_{i,y}(t) \rangle$$

acceleration of particle i at time t

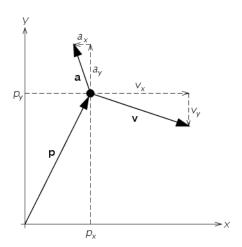
$$\vec{a}_i(t) = \langle a_{i,x}(t), a_{i,y}(t) \rangle$$

▶ force acting on particle *i* at time *t*

$$\vec{F}_i(t) = \langle F_{i,x}(t), F_{i,y}(t) \rangle$$

N-body Problems: Newtonian Physics

Physical characteristics:



N-body Problems: Newtonian Physics

Motion and forces:

velocity is the time derivative of position

$$\vec{v}_i(t) = \frac{d}{dt}\vec{p}_i(t)$$

- acceleration is the time derivative of velocity
- acceleration is the second time derivative of position

$$\vec{a}_i(t) = \frac{d}{dt}\vec{v}_i(t) = \frac{d^2}{dt^2}\vec{p}_i(t)$$

Newton's Second Law of Motion

$$\vec{F}_i(t) = m \cdot \vec{a}_i(t)$$
 $\vec{a}_i(t) = \frac{1}{m} \cdot \vec{F}_i(t)$

A particle's position at time $t + \delta$ is related to its position at time t according to a Taylor series expansion:

$$\vec{p}_i(t+\delta) = \vec{p}_i(t) + \delta \cdot \frac{d}{dt}\vec{p}_i(t) + \frac{\delta^2}{2} \cdot \frac{d^2}{dt^2}\vec{p}_i(t) + O(\delta^3)$$

A particle's position at time $t+\delta$ is related to its position at time t according to a Taylor series expansion:

$$\vec{p}_i(t+\delta) = \vec{p}_i(t) + \delta \cdot \frac{d}{dt}\vec{p}_i(t) + \frac{\delta^2}{2} \cdot \frac{d^2}{dt^2}\vec{p}_i(t) + O(\delta^3)$$

If δ is small, then $O(\delta^3)$ is neglible and a particle's future position may be approximated:

$$\vec{p}_i(t+\delta) \approx \vec{p}_i(t) + \delta \cdot \vec{v}_i(t) + \frac{\delta^2}{2} \cdot \vec{a}_i(t)$$

A particle's velocity at time $t+\delta$ is related to its velocity at time t according to a Taylor series expansion:

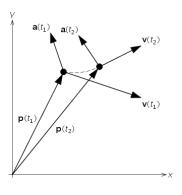
$$\vec{v}_i(t+\delta) = \vec{v}_i(t) + \delta \cdot \frac{d}{dt}\vec{v}_i(t) + O(\delta^2)$$

A particle's velocity at time $t+\delta$ is related to its velocity at time t according to a Taylor series expansion:

$$\vec{v}_i(t+\delta) = \vec{v}_i(t) + \delta \cdot \frac{d}{dt}\vec{v}_i(t) + O(\delta^2)$$

If δ is small, then $O(\delta^2)$ is neglible and a particle's future velocity may be approximated:

$$\vec{v}_i(t+\delta) \approx \vec{v}_i(t) + \delta \cdot \vec{a}_i(t)$$



$$t_2 = t_1 + \delta$$

$$\vec{p}_i(t_2) = \vec{p}_i(t_1) + \delta \cdot \vec{v}_i(t_1) + \frac{\delta^2}{2} \cdot \vec{a}_i(t_1)$$

$$\vec{v}_i(t_2) = \vec{v}_i(t_1) + \delta \cdot \vec{a}_i(t_1)$$

10

Suppose we know the positions and velocities of n particles at time 0, and wish to predict their motion into the future.

Suppose we know the positions and velocities of n particles at time 0, and wish to predict their motion into the future.

```
for i=0 to n-1
\vec{p}_i \leftarrow \text{Position of particle } i at time 0
\vec{v}_i \leftarrow \text{Velocity of particle } i at time 0
for k=1 to steps
for i=0 to n-1
\vec{F} \leftarrow \text{Calculate force on particle } i \text{ at time } (k-1)\delta
from external fields and the positions and velocities of all particles \vec{a}_i \leftarrow \frac{1}{m_i} \cdot \vec{F} // Acceleration of particle i at time (k-1)\delta
for i=0 to n-1
\vec{p}_i \leftarrow \vec{p}_i + \delta \vec{v}_i + 0.5\delta^2 \vec{a}_i // Position of particle i at time k\delta
\vec{v}_i \leftarrow \vec{v}_i + \delta \vec{a}_i // Velocity of particle i at time k\delta
```

The previous is an example of a *numerical integration* algorithm:

- Solves the differential equation for $\vec{p}_i(t)$ via a sequence of discrete time-step calculations
- ▶ Uses *steps* time steps
- Uses a second-order integration algorithm (for positions)
 - lacktriangle Uses terms of order up to and including δ^2
- ▶ Incurs *truncation errors* due to the Taylor-series approximation
 - ightharpoonup For accurate motion, requires a very small δ and a very large **steps**

The previous is an example of a *numerical integration* algorithm:

- ▶ Solves the differential equation for $\vec{p}_i(t)$ via a sequence of discrete time-step calculations
- Uses steps time steps
- Uses a second-order integration algorithm (for positions)
 - lacktriangle Uses terms of order up to and including δ^2
- ▶ Incurs *truncation errors* due to the Taylor-series approximation
 - ightharpoonup For accurate motion, requires a very small δ and a very large **steps**

Note: Only need the *last* positions and velocities of particles to calculate the *next* positions and velocities (like cellular automata algorithms).

The previous is an example of a *numerical integration* algorithm:

- ▶ Solves the differential equation for $\vec{p}_i(t)$ via a sequence of discrete time-step calculations
- Uses steps time steps
- Uses a second-order integration algorithm (for positions)
 - lacktriangle Uses terms of order up to and including δ^2
- ▶ Incurs *truncation errors* due to the Taylor-series approximation
 - ightharpoonup For accurate motion, requires a very small δ and a very large **steps**

Note: Only need the *last* positions and velocities of particles to calculate the *next* positions and velocities (like cellular automata algorithms).

But, to visualize the particles' motion, need to record particles' positions at different times.

The previous is an example of a *numerical integration* algorithm:

- ▶ Solves the differential equation for $\vec{p}_i(t)$ via a sequence of discrete time-step calculations
- Uses steps time steps
- Uses a second-order integration algorithm (for positions)
 - lacktriangle Uses terms of order up to and including δ^2
- ▶ Incurs *truncation errors* due to the Taylor-series approximation
 - ightharpoonup For accurate motion, requires a very small δ and a very large **steps**

Note: Only need the *last* positions and velocities of particles to calculate the *next* positions and velocities (like cellular automata algorithms).

But, to visualize the particles' motion, need to record particles' positions at different times.

With a very small δ , don't need to record particles' positions at *each* time; only need to record a "snapshot" of positions after a number of time steps.

Only need to record a "snapshot" of positions after a number of time steps.

Only need to record a "snapshot" of positions after a number of time steps.

```
for i = 0 to n - 1
        \vec{p}_i \leftarrow \text{Position of particle } i \text{ at time } 0
        \vec{v}_i \leftarrow \text{Velocity of particle } i \text{ at time } 0
writeSnapshot (\vec{p})
for s = 1 to snaps
        for k = 1 to steps
                for i = 0 to n - 1
                         \vec{F} \leftarrow \text{Calculate force on particle } i \text{ at time } ((s-1)steps + (k-1))\delta \text{ from external fields and the positions and velocities of all particles}
                         \vec{a}_i \leftarrow \frac{1}{m} \cdot \tilde{F} // Acceleration of particle i at time ((s-1)steps+(k-1))\delta
                for i = 0 to n - 1
                         \vec{p}_i \leftarrow \vec{p}_i + \delta \vec{v}_i + 0.5\delta^2 \vec{a}_i // Position of particle i at time ((s-1)steps+k)\delta
                         \vec{v}_i \leftarrow \vec{v}_i + \delta \vec{a}_i // Velocity of particle i at time ((s-1)steps+k)\delta
        writeSnapshot(p)
```

The previous is an example of a *numerical integration* algorithm:

- ▶ Solves the differential equation for $\vec{p}_i(t)$ via a sequence of discrete time-step calculations
- Uses steps time steps
- Uses a second-order integration algorithm (for positions)
 - lacktriangle Uses terms of order up to and including δ^2
- ▶ Incurs *truncation errors* due to the Taylor-series approximation
 - lacktriangleright For accurate motion, requires a very small δ and a very large s

Accuracy degrades due to truncation errors and floating-point arithmetic; need a mechanism to monitor accuracy.

The previous is an example of a *numerical integration* algorithm:

- Solves the differential equation for $\vec{p}_i(t)$ via a sequence of discrete time-step calculations
- Uses steps time steps
- Uses a second-order integration algorithm (for positions)
 - lacktriangle Uses terms of order up to and including δ^2
- ▶ Incurs *truncation errors* due to the Taylor-series approximation
 - lacktriangle For accurate motion, requires a very small δ and a very large s

Accuracy degrades due to truncation errors and floating-point arithmetic; need a mechanism to monitor accuracy.

Keep track of a conserved quantity:

- momentum Law of Conservation of Momentum
- energy Law of Conservation of Energy

Such a quantity should remain constant.

Keep track of a conserved quantity.

Keep track of a conserved quantity.

```
for i = 0 to n - 1
        \vec{p}_i \leftarrow \text{Position of particle } i \text{ at time } 0
         \vec{v}_i \leftarrow \text{Velocity of particle } i \text{ at time } 0
cq 

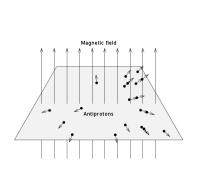
Calculate conserved quantity at time 0 from positions and velocities of all particles
writeSnapshot(p, cq)
for s = 1 to snaps
        for k = 1 to steps
                  for i = 0 to n - 1
                          \vec{F} \leftarrow \text{Calculate force on particle } i \text{ at time } ((s-1)steps+(k-1))\delta \text{ from external fields and the positions and velocities of all particles}
                          \vec{a}_i \leftarrow \frac{1}{m} \cdot \vec{F} // Acceleration of particle i at time ((s-1)steps+(k-1))\delta
                  for i = 0 to n - 1
                          \vec{p}_i \leftarrow \vec{p}_i + \delta \vec{v}_i + 0.5\delta^2 \vec{a}_i // Position of particle i at time ((s-1)steps+k)\delta
                           \vec{v}_i \leftarrow \vec{v}_i + \delta \vec{a}_i // Velocity of particle i at time ((s-1)steps+k)\delta
        cq \leftarrow \frac{\text{Calculate conserved quantity at time } ((s-1)steps+k)\delta}{\text{from positions and velocities of all particles}}
         writeSnapshot (\vec{p}, cq)
```

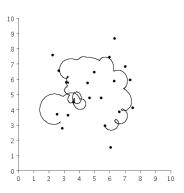
15

N-body Problems: Antiproton Motion

Electro-magnetic **N**-body problems:

motion of charged particles (protons, antiprotons),
 due to mutual attraction/repulsion and magnetic fields





16

Antiproton Motion

Physical characteristics:

charge of particle i

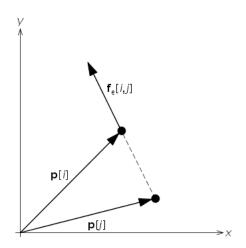
 q_i

force acting on particle i at time t

$$\vec{F}_i(t) = \sum_{j \neq i} \vec{F}_{e,i,j}(t) + \vec{F}_{m,i}(t)$$

- $ightharpoonup \vec{F}_{e,i,j}(t)$: electrostatic force acting on particle i at time t from particle j
- $ightharpoonup \vec{F}_{m,i}(t)$: magnetic force acting on particle i at time t due to magnetic field
- ▶ total linear momentum of all particles at time t (conserved quantity)

$$\vec{P}(t) = \sum_{i} m_{i} \vec{v}_{i}(t)$$



$$\vec{F}_{e,i,j}(t) \propto \vec{p}_i(t) - \vec{p}_j(t)$$

$$ec{F}_{e,i,j}(t) \propto rac{ec{p}_i(t) - ec{p}_j(t)}{|ec{p}_i(t) - ec{p}_j(t)|}$$

$$ec{F}_{e,i,j}(t) \propto ec{p}_i(t) - ec{p}_j(t)$$

$$ec{F}_{e,i,j}(t) \propto rac{ec{p}_i(t) - ec{p}_j(t)}{|ec{p}_i(t) - ec{p}_j(t)|}$$

$$\vec{F}_{e,i,j}(t) = k_e \frac{q_i q_j}{|\vec{p}_i(t) - \vec{p}_j(t)|^2} \cdot \frac{\vec{p}_i(t) - \vec{p}_j(t)}{|\vec{p}_i(t) - \vec{p}_j(t)|}$$

$$ec{F}_{e,i,j}(t) \propto ec{p}_i(t) - ec{p}_j(t)$$

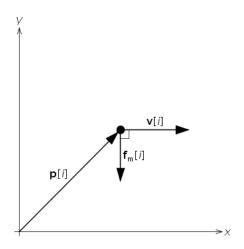
$$ec{F}_{e,i,j}(t) \propto rac{ec{p}_i(t) - ec{p}_j(t)}{|ec{p}_i(t) - ec{p}_j(t)|}$$

$$\vec{F}_{e,i,j}(t) = k_e \frac{q_i q_j}{|\vec{p}_i(t) - \vec{p}_j(t)|^2} \cdot \frac{\vec{p}_i(t) - \vec{p}_j(t)}{|\vec{p}_i(t) - \vec{p}_j(t)|}$$

$$\vec{F}_{e,i,j}(t) = rac{k_e q_i q_j(\vec{p}_i(t) - \vec{p}_j(t))}{|\vec{p}_i(t) - \vec{p}_i(t)|^3}$$

Antiproton Motion: Lorentz Force

 $ec{F}_{m,i}(t)$: magnetic force acting on particle i at time t due to magnetic field



Antiproton Motion: Lorentz Force

 $ec{F}_{m,i}(t)$: magnetic force acting on particle i at time t due to magnetic field

$$\vec{F}_{m,i}(t) \propto \langle v_{i,y}(t), -v_{i,x}(t) \rangle$$

$$ec{F}_{m,i}(t) \propto rac{\langle v_{i,y}(t), -v_{i,x}(t)
angle}{|ec{v}_i(t)|}$$

Antiproton Motion: Lorentz Force

 $ec{F}_{m,i}(t)$: magnetic force acting on particle i at time t due to magnetic field

$$ec{F}_{m,i}(t) \propto \langle v_{i,y}(t), -v_{i,x}(t) \rangle$$
 $ec{F}_{m,i}(t) \propto rac{\langle v_{i,y}(t), -v_{i,x}(t) \rangle}{|ec{v}_i(t)|}$

$$\vec{F}_{m,i}(t) = (q_i|\vec{v}_i(t)|B_m) \cdot \frac{\langle v_{i,y}(t), -v_{i,x}(t) \rangle}{|\vec{v}_i(t)|}$$

Antiproton Motion: Lorentz Force

 $ec{F}_{m,i}(t)$: magnetic force acting on particle i at time t due to magnetic field

$$ec{F}_{m,i}(t) \propto \langle v_{i,y}(t), -v_{i,x}(t)
angle \ ec{F}_{m,i}(t) \propto rac{\langle v_{i,y}(t), -v_{i,x}(t)
angle}{|ec{v}_i(t)|}$$

$$\vec{F}_{m,i}(t) = (q_i | \vec{v}_i(t) | B_m) \cdot \frac{\langle v_{i,y}(t), -v_{i,x}(t) \rangle}{|\vec{v}_i(t)|}$$

$$\vec{F}_{m,i}(t) = q_i B_m \langle v_{i,y}(t), -v_{i,x}(t) \rangle$$

```
for i = 0 to n - 1
         \vec{p}_i \leftarrow \text{Position of particle } i \text{ at time } 0
         \vec{v}_i \leftarrow \text{Velocity of particle } i \text{ at time } 0
for i = 0 to n - 1
         \vec{P} \leftarrow \vec{P} + m_i \vec{v}_i
writeSnapshot (\vec{p}, \vec{P})
for s = 1 to snaps
         for k = 1 to steps
                   for i = 0 to n - 1
                             \vec{F} \leftarrow \vec{0}
                             for j = 0 to n - 1, j \neq i
                                    \vec{d} \leftarrow \vec{p}_i(t) - \vec{p}_j(t) 
\vec{F} \leftarrow \vec{F} + k_e q_i q_j \cdot \vec{d}/|\vec{d}|^3
                            \vec{F} \leftarrow \vec{F} + q_i B_m \cdot \langle v_{i,y}, -v_{i,x} \rangle
                   for i = 0 to n - 1
                            \begin{array}{lll} \vec{p}_i & \leftarrow & \vec{p}_i + \delta \vec{v}_i + 0.5 \delta^2 \vec{a}_i \\ \vec{v}_i & \leftarrow & \vec{v}_i + \delta \vec{a}_i \end{array}
         for i = 0 to n - 1
                   \vec{P} \leftarrow \vec{P} + m_i \vec{v}_i
         writeSnapshot (\vec{p}, \vec{P})
```

```
for i = 0 to n - 1
         \vec{p}_i \leftarrow \text{Position of particle } i \text{ at time } 0
         \vec{v}_i \leftarrow \text{Velocity of particle } i \text{ at time } 0
for i = 0 to n - 1
         \vec{P} \leftarrow \vec{P} + m: \vec{v}:
writeSnapshot (\vec{p}, \vec{P})
for s = 1 to snaps
         for k = 1 to steps
                  for i = 0 to n - 1
                           \vec{a}_i \leftarrow \vec{0}
                           for j = 0 to n - 1, j \neq i
                                     \vec{d} \leftarrow \vec{p}_i(t) - \vec{p}_i(t)
                                    \vec{a}_i \leftarrow \vec{a}_i + \frac{1}{m_i} \cdot k_e q_i q_j \cdot \vec{d} / |\vec{d}|^3
                           \vec{a}_i \leftarrow \vec{a}_i + \frac{1}{m} \cdot q_i B_m \cdot \langle v_{i,y}, -v_{i,x} \rangle
                  for i = 0 to n - 1
                          \begin{array}{ll} \vec{p}_i \ \leftarrow \ \vec{p}_i + \delta \vec{v}_i + 0.5 \delta^2 \vec{a}_i \\ \vec{v}_i \ \leftarrow \ \vec{v}_i + \delta \vec{a}_i \end{array}
         for i = 0 to n - 1
                  \vec{P} \leftarrow \vec{P} + m_i \vec{v}_i
         writeSnapshot (\vec{p}, \vec{P})
```

Calculate acceleration directly.

```
for i = 0 to n - 1
         \vec{p}_i \leftarrow \text{Position of particle } i \text{ at time } 0
         \vec{v}_i \leftarrow \text{Velocity of particle } i \text{ at time } 0
for i = 0 to n - 1
         \vec{P} \leftarrow \vec{P} + 1 \cdot \vec{v}
writeSnapshot (\vec{p}, \vec{P})
for s = 1 to snaps
         for k = 1 to steps
                   for i = 0 to n - 1
                            for j = 0 to n - 1, j \neq i
                                      \vec{d} \leftarrow \vec{p}_i(t) - \vec{p}_i(t)
                                     \vec{a}_i \leftarrow \vec{a}_i + Q \cdot \vec{d}/|\vec{d}|^3
                  \vec{a}_i \leftarrow \vec{a}_i + \vec{B} \cdot \langle v_{i,y}, -v_{i,x} \rangle
for i = 0 to n - 1
                           \begin{array}{ll} \vec{p}_i & \leftarrow & \vec{p}_i + \delta \vec{v}_i + 0.5 \delta^2 \vec{a}_i \\ \vec{v}_i & \leftarrow & \vec{v}_i + \delta \vec{a}_i \end{array}
         for i = 0 to n - 1
                   \vec{P} \leftarrow \vec{P} + 1 \cdot \vec{v}_i
         writeSnapshot (\vec{p}, \vec{P})
```

Assume all particles have unit mass and unit charge (by adjusting units and introducing constants Q and B).

```
for i = 0 to n - 1
        \vec{p}_i \leftarrow \langle \text{random}(R/4, 3R/4), \text{random}(R/4, 3R/4) \rangle
\vec{P} \leftarrow \vec{0}
writeSnapshot (\vec{p}, \vec{P})
for s = 1 to snaps
         for k = 1 to steps
                  for i = 0 to n - 1
                           \vec{a}: \leftarrow \vec{0}
                           for i = 0 to n - 1, i \neq i
                            \vec{d} \leftarrow \vec{p}_i(t) - \vec{p}_j(t) 
 \vec{a}_i \leftarrow \vec{a}_i + Q \cdot \vec{d}/|\vec{d}|^3 
 \vec{a}_i \leftarrow \vec{a}_i + B \cdot \langle v_{i,y}, -v_{i,x} \rangle 
                  for i = 0 to n - 1
                           \vec{p}_i \leftarrow \vec{p}_i + \delta \vec{v}_i + 0.5 \delta^2 \vec{a}_i
         for i = 0 to n - 1
                  \vec{P} \leftarrow \vec{P} + 1 \cdot \vec{v}_i
         writeSnapshot (\vec{p}, \vec{P})
```

Initial position is random (near center of the square (0,0) to (R,R)). Initial velocity is $\vec{0}$.

```
for i = 0 to n - 1
        \vec{p}_i \leftarrow \langle \text{random}(R/4, 3R/4), \text{random}(R/4, 3R/4) \rangle
writeSnapshot (\vec{p}, \vec{P})
for s = 1 to snaps
        for k = 1 to steps
               for i = 0 to n - 1
                       for i = 0 to n - 1, i \neq i
                               \vec{d} \leftarrow \vec{p}_i(t) - \vec{p}_i(t)
                               \vec{a}_i \ \leftarrow \ \vec{a}_i + Q \cdot \vec{d}/|\vec{d}|^3
               for i = 0 to n - 1
                       \vec{a}_i \leftarrow \vec{a}_i + B \cdot \langle v_{i,y}, -v_{i,x} \rangle
                       \vec{p}_i \leftarrow \vec{p}_i + \delta \vec{v}_i + 0.5 \delta^2 \vec{a}_i
        for i = 0 to n - 1
                \vec{P} \leftarrow \vec{P} + 1 \cdot \vec{v}_i
        writeSnapshot (\vec{p}, \vec{P})
```

Rearrange acceleration computation.

AntiprotonSeq.java

code/AntiprotonSeq.java

- seed PRNG seed for particles' initial positions
- ▶ *R* side of the square for particles' initial positions
- dt size of the time step, δ
- steps number of time steps in each snapshot
- snaps numer of snapshots
- ▶ n number of particles
- outfile output file name
- edu.rit.clu.antimatter.AntiprotonFile handles file I/O of snapshots
 - edu.rit.clu.antimatter.AntiprotonPlot produces visualizations
- edu.rit.vector.Vector2D vector arithmetic

How can we parallelize this algorithm for a cluster parallel computer?

```
for i = 0 to n - 1
          \vec{p}_i \leftarrow \langle \text{random}(R/4, 3R/4), \text{random}(R/4, 3R/4) \rangle
writeSnapshot (\vec{p}, \vec{P})
for s = 1 to snaps
          for k = 1 to steps
                    for i = 0 to n - 1
                               for i = 0 to n - 1, i \neq i
                                         \vec{d} \leftarrow \vec{p}_i(t) - \vec{p}_j(t)

\vec{a}_i \leftarrow \vec{a}_i + Q \cdot \vec{d}/|\vec{d}|^3
                              \vec{a}_{i} \leftarrow \vec{a}_{i} + B \cdot \langle v_{i,y}, -v_{i,x} \rangle
\vec{p}_{i} \leftarrow \vec{p}_{i} + \delta \vec{v}_{i} + 0.5\delta^{2} \vec{a}_{i}
\vec{v}_{i} \leftarrow \vec{v}_{i} + \delta \vec{a}_{i}
          for i = 0 to n - 1
                     \vec{P} \leftarrow \vec{P} + 1 \cdot \vec{v}
          writeSnapshot (\vec{p}, \vec{P})
```

How can we parallelize this algorithm for a cluster parallel computer?

```
for i = 0 to n - 1
         \vec{p}_i \leftarrow \langle \text{random}(R/4, 3R/4), \text{random}(R/4, 3R/4) \rangle
writeSnapshot (\vec{p}, \vec{P})
for s = 1 to snaps
         for k = 1 to steps
                   for i = 0 to n - 1
                             for i = 0 to n - 1, i \neq i
                                       \vec{d} \leftarrow \vec{p}_i(t) - \vec{p}_i(t)
                                      \vec{a}_i \leftarrow \vec{a}_i + Q \cdot \vec{d}/|\vec{d}|^3
                            \vec{a}_{i} \leftarrow \vec{a}_{i} + B \cdot \langle v_{i,y}, -v_{i,x} \rangle
\vec{p}_{i} \leftarrow \vec{p}_{i} + \delta \vec{v}_{i} + 0.5\delta^{2} \vec{a}_{i}
\vec{v}_{i} \leftarrow \vec{v}_{i} + \delta \vec{a}_{i}
         for i = 0 to n - 1
                   \vec{P} \leftarrow \vec{P} + 1 \cdot \vec{v}_i
         writeSnapshot (\vec{p}, \vec{P})
```

Partition the n particles among the K processes; each process responsible for n/K particles (position, velocity, acceleration).

Partition the n particles among the K processes; each process responsible for n/K particles (position, velocity, acceleration).

```
lb, ub \leftarrow range(0, n-1, K, rank)
for i = lb to ub
       \vec{p}_i \leftarrow \langle \text{random}(R/4, 3R/4), \text{random}(R/4, 3R/4) \rangle
writeSnapshot (\vec{p}, \vec{P})
for s = 1 to snaps
        for k = 1 to steps
                for i = lb to ub
                        for i = 0 to n - 1, i \neq i
                                 \vec{d} \leftarrow \vec{p}_i(t) - \vec{p}_i(t)
                                 \vec{a}_i \leftarrow \vec{a}_i + Q \cdot \vec{d}/|\vec{d}|^3
                for i = lb to ub
                        \vec{a}_i \leftarrow \vec{a}_i + B \cdot \langle v_{i,y}, -v_{i,x} \rangle
                        \vec{p}_i \leftarrow \vec{p}_i + \delta \vec{v}_i + 0.5\delta^2 \vec{a}_i
\vec{v}_i \leftarrow \vec{v}_i + \delta \vec{a}_i
        for i = lb to ub
                \vec{P} \leftarrow \vec{P} + 1 \cdot \vec{v}
        writeSnapshot (\vec{p}, \vec{P})
```

Partition the n particles among the K processes; each process responsible for n/K particles (position, velocity, acceleration).

```
lb, ub \leftarrow range(0, n-1, K, rank)
for i = lb to ub
        \vec{p}_i \leftarrow \langle \text{random}(R/4, 3R/4), \text{random}(R/4, 3R/4) \rangle
writeSnapshot (\vec{p}, \vec{P})
for s = 1 to snaps
        for k = 1 to steps
                for i = lb to ub
                        for i = 0 to n - 1, i \neq i
                                 \vec{d} \leftarrow \vec{p}_i(t) - \vec{p}_i(t)
                                 \vec{a}_i \leftarrow \vec{a}_i + Q \cdot \vec{d}/|\vec{d}|^3
                for i = lb to ub
                        \vec{a}_i \leftarrow \vec{a}_i + B \cdot \langle v_{i,y}, -v_{i,x} \rangle
                        \vec{p}_i \leftarrow \vec{p}_i + \delta \vec{v}_i + 0.5\delta^2 \vec{a}_i
\vec{v}_i \leftarrow \vec{v}_i + \delta \vec{a}_i
        for i = lb to ub
                \vec{P} \leftarrow \vec{P} + 1 \cdot \vec{v}_i
        writeSnapshot (\vec{p}, \vec{P})
```

Note: use parallel output files pattern.

Partition the n particles among the K processes; each process responsible for n/K particles (position, velocity, acceleration).

$$\begin{array}{ll} \text{for } i = lb \text{ to } ub \\ \text{for } j = 0 \text{ to } n-1, \ j \neq i \\ \vec{d} \leftarrow \vec{p_i}(t) - \vec{p_j}(t) \\ \vec{a_i} \leftarrow \vec{a_i} + Q \cdot \vec{d}/|\vec{d}|^3 \end{array}$$

$$\begin{array}{ll} \text{for } i = lb \text{ to } ub \\ \vec{a_i} \leftarrow \vec{a_i} + B \cdot \langle v_{i,y}, -v_{i,x} \rangle \\ \vec{p_i} \leftarrow \vec{p_j} + \delta \vec{v_i} + 0.5 \delta^2 \vec{a_i} \\ \vec{v_i} \leftarrow \vec{v_i} + \delta \vec{a_i} \\ \vec{a_i} \leftarrow \vec{0} \end{array}$$

What data is accessed by each process on each iteration?

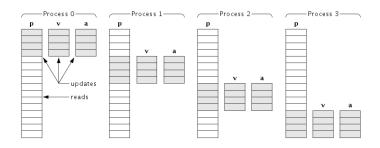
Partition the n particles among the K processes; each process responsible for n/K particles (position, velocity, acceleration).

$$\begin{array}{ll} \text{for } i = lb \text{ to } ub \\ \text{for } j = 0 \text{ to } n-1, \ j \neq i \\ \overrightarrow{d} \leftarrow \overrightarrow{p_i}(t) - \overrightarrow{p_j}(t) \\ \overrightarrow{a_i} \leftarrow \overrightarrow{a_i} + Q \cdot \overrightarrow{d}/|\overrightarrow{d}|^3 \end{array}$$

$$\begin{array}{ll} \text{for } i = lb \text{ to } ub \\ \overrightarrow{a_i} \leftarrow \overrightarrow{a_i} + B \cdot \langle v_{i,y}, -v_{i,x} \rangle \\ \overrightarrow{p_i} \leftarrow \overrightarrow{p_i} + \delta \overrightarrow{v_i} + 0.5\delta^2 \overrightarrow{a_i} \\ \overrightarrow{v_i} \leftarrow \overrightarrow{v_i} + \delta \overrightarrow{a_i} \\ \overrightarrow{a_i} \leftarrow \overrightarrow{0} \end{array}$$

What data is accessed by each process on each iteration?

Every process updates its slice of position, velocity, and acceleration arrays. But, every process reads the entire position array.



Each process need only allocate storage for its slice of velocity and acceleration arrays, but must allocate storage for entire position array.

Partition the n particles among the K processes; each process responsible for n/K particles (position, velocity, acceleration).

$$\begin{array}{ll} \text{for } i = lb \text{ to } ub \\ \text{for } j = 0 \text{ to } n-1, \ j \neq i \\ \overrightarrow{d} \leftarrow \overrightarrow{p_i}(t) - \overrightarrow{p_j}(t) \\ \overrightarrow{a_i} \leftarrow \overrightarrow{a_i} + Q \cdot \overrightarrow{d}/|\overrightarrow{d}|^3 \end{array}$$

$$\begin{array}{ll} \text{for } i = lb \text{ to } ub \\ \overrightarrow{a_i} \leftarrow \overrightarrow{a_i} + B \cdot \langle v_{i,y}, -v_{i,x} \rangle \\ \overrightarrow{p_i} \leftarrow \overrightarrow{p_i} + \delta \overrightarrow{v_i} + 0.5\delta^2 \overrightarrow{a_i} \\ \overrightarrow{v_i} \leftarrow \overrightarrow{v_i} + \delta \overrightarrow{a_i} \\ \overrightarrow{a_i} \leftarrow \overrightarrow{0} \end{array}$$

Partition the n particles among the K processes; each process responsible for n/K particles (position, velocity, acceleration).

$$\begin{array}{ll} \text{for } i = lb \text{ to } ub \\ \text{ for } j = 0 \text{ to } n-1, \ j \neq i \\ \overrightarrow{d} \leftarrow \overrightarrow{p_i}(t) - \overrightarrow{p_j}(t) \\ \overrightarrow{a_i} \leftarrow \overrightarrow{a_i} + Q \cdot \overrightarrow{d} / |\overrightarrow{d}|^3 \end{array}$$

$$\text{for } i = lb \text{ to } ub \\ \overrightarrow{a_i} \leftarrow \overrightarrow{a_i} + B \cdot \langle v_{i,y}, -v_{i,x} \rangle \\ \overrightarrow{p_i} \leftarrow \overrightarrow{p_i} + \delta \overrightarrow{v_i} + 0.5\delta^2 \overrightarrow{a_i} \\ \overrightarrow{v_i} \leftarrow \overrightarrow{v_i} + \delta \overrightarrow{a_i} \\ \overrightarrow{a_i} \leftarrow \overrightarrow{0} \end{array}$$

Every process updates its slice of position, velocity, and acceleration arrays. But, every process reads the entire position array.

After updating its slice of position and velocity, but before calculating its next slice of acceleration, every process must communicate its slice of position to every other process.

Which collective communication operation?

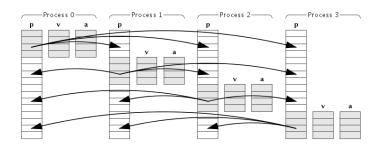
Partition the n particles among the K processes; each process responsible for n/K particles (position, velocity, acceleration).

$$\begin{array}{ll} \text{for } i = lb \text{ to } ub \\ \text{ for } j = 0 \text{ to } n-1, \ j \neq i \\ \overrightarrow{d} \leftarrow \overrightarrow{p_i}(t) - \overrightarrow{p_j}(t) \\ \overrightarrow{a_i} \leftarrow \overrightarrow{a_i} + Q \cdot \overrightarrow{d}/|\overrightarrow{d}|^3 \\ \text{for } i = lb \text{ to } ub \\ \overrightarrow{a_i} \leftarrow \overrightarrow{a_i} + B \cdot \langle v_{i,y}, -v_{i,x} \rangle \\ \overrightarrow{p_i} \leftarrow \overrightarrow{p_i} + \delta \overrightarrow{v_i} + 0.5 \delta^2 \overrightarrow{a_i} \\ \overrightarrow{v_i} \leftarrow \overrightarrow{v_i} + \delta \overrightarrow{a_i} \\ \overrightarrow{a_i} \leftarrow \overrightarrow{0} \end{array}$$

Every process updates its slice of position, velocity, and acceleration arrays. But, every process reads the entire position array.

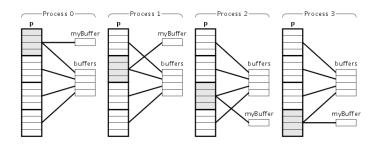
After updating its slice of position and velocity, but before calculating its next slice of acceleration, every process must communicate its slice of position to every other process.

Which collective communication operation? all-gather

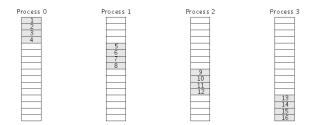


After updating its slice of position and velocity, but before calculating its next slice of acceleration, every process must communicate its slice of position to every other process.

Also acts as a synchronization point to enforce the sequential dependency from one time step to the next.



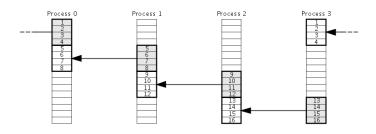
world.allGather (myBuffer, buffers);



Each node simultaneously sends one data buffer to its predecessor and receives another data buffer from its successor.

First round, sends its own data buffer;

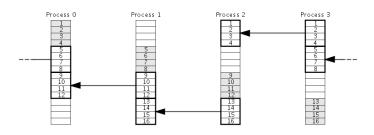
later rounds, sends last received data buffer.



Each node simultaneously sends one data buffer to its predecessor and receives another data buffer from its successor.

First round, sends its own data buffer;

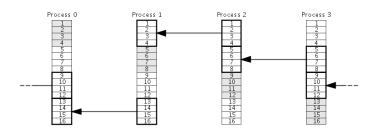
later rounds, sends last received data buffer.



Each node simultaneously sends one data buffer to its predecessor and receives another data buffer from its successor.

First round, sends its own data buffer;

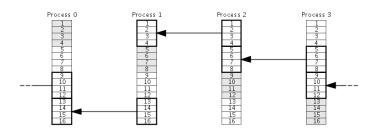
later rounds, sends last received data buffer.



Each node simultaneously sends one data buffer to its predecessor and receives another data buffer from its successor.

First round, sends its own data buffer;

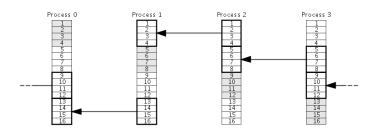
later rounds, sends last received data buffer.



Each node simultaneously sends one data buffer to its predecessor and receives another data buffer from its successor.

First round, sends its own data buffer; later rounds, sends last received data buffer.

$$T_{\text{all-gather}}(b, K) =$$



Each node simultaneously sends one data buffer to its predecessor and receives another data buffer from its successor.

First round, sends its own data buffer; later rounds, sends last received data buffer.

$$T_{\text{all-gather}}(b,K) = (L + \frac{1}{B}b)(K-1)$$

Note: Assumes same latency and bandwith for both ${\bf 1}$ send and ${\bf K}$ simultaneous sends.

AntiprotonClu.java

code/AntiprotonClu.java

AntiprotonClu.java

code/AntiprotonClu.java

Before looking at the running-time measurements for AntiprotonClu, derive a model to predict the running time.

What will we need?

- a calculation time model
- ▶ a communication time model

Calculation Time Model

Antiproton motion algorithm is $O(sn^2)$, where n is the number of particles and s is total number of steps $(s = snaps \cdot steps)$.

$$T_{\rm calc}^{\bar{p}}(s,n,1) = asn^2$$

$$T_{\mathrm{calc}}^{\bar{p}}(s,n,K) = asn^2 \frac{1}{K}$$

$$T_{\rm calc}^{\bar{p}}(s,n,1) = 2.04 \times 10^{-8} sn^2$$

$$T_{\mathrm{calc}}^{\bar{p}}(s,n,K) = 2.04 \times 10^{-8} sn^2 \frac{1}{K}$$

Communication Time Model

- ► How many all-gathers are performed?
- ▶ How many bits in each all-gather message?

Communication Time Model

- ► How many all-gathers are performed? s
- ► How many bits in each all-gather message? 128 $n\frac{1}{K}$

Communication Time Model

- How many all-gathers are performed?
- ► How many bits in each all-gather message? 128 $n\frac{1}{K}$

$$T_{\text{comm}}^{\bar{p}}(s, n, K) = sT_{\text{all-gather}}(128n\frac{1}{K}, K)$$

$$T_{\text{comm}}^{\bar{p}}(s, n, K) = s(L + \frac{1}{B}128n\frac{1}{K})(K - 1)$$

$$T_{\text{comm}}^{\bar{p}}(s, n, K) = s(2.08 \times 10^{-4} + 1.37 \times 10^{-7}n\frac{1}{K})(K - 1)$$

$$T^{\bar{p}}(s,n,K) = T_{\mathrm{calc}}^{\bar{p}}(s,n,K) + T_{\mathrm{comm}}^{\bar{p}}(s,n,K)$$

$$T^{\bar{p}}(s, n, K) = asn^2 \frac{1}{K} + s(L + \frac{1}{B}128n\frac{1}{K})(K - 1)$$

$$T^{\bar{p}}(s, n, K) = 2.04 \times 10^{-8} sn^2 \frac{1}{K} + s(2.08 \times 10^{-4} + 1.37 \times 10^{-7} n \frac{1}{K})(K - 1)$$

$$T^{\bar{p}}(s,n,K) = T_{\mathrm{calc}}^{\bar{p}}(s,n,K) + T_{\mathrm{comm}}^{\bar{p}}(s,n,K)$$

$$T^{\bar{p}}(s, n, K) = asn^2 \frac{1}{K} + s(L + \frac{1}{B}128n\frac{1}{K})(K - 1)$$

$$T^{\bar{p}}(s, n, K) = 2.04 \times 10^{-8} sn^2 \frac{1}{K} + s(2.08 \times 10^{-4} + 1.37 \times 10^{-7} n \frac{1}{K})(K - 1)$$

Classic cluster parallel program problem:

- calculation time decreases as a function of K
- communication time increases as a function of K

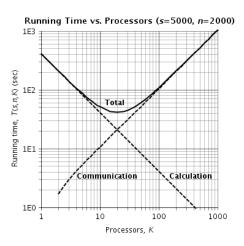
$$T^{\bar{p}}(s,n,K) = T_{\mathrm{calc}}^{\bar{p}}(s,n,K) + T_{\mathrm{comm}}^{\bar{p}}(s,n,K)$$

$$T^{\bar{p}}(s, n, K) = asn^2 \frac{1}{K} + s(L + \frac{1}{B}128n\frac{1}{K})(K - 1)$$

$$T^{\bar{p}}(s, n, K) = 2.04 \times 10^{-8} sn^2 \frac{1}{K} + s(2.08 \times 10^{-4} + 1.37 \times 10^{-7} n \frac{1}{K})(K - 1)$$

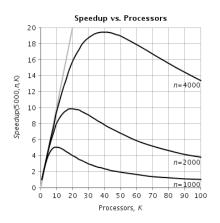
Classic cluster parallel program problem:

- calculation time decreases as a function of K
- communication time increases as a function of K
 - but now O(K), rather than only $O(\log_2 K)$

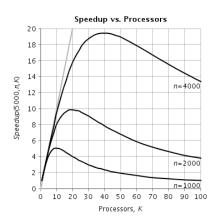


$$T^{\bar{p}}(s,n,K) = 2.04 \times 10^{-8} sn^2 \frac{1}{K} + s(2.08 \times 10^{-4} + 1.37 \times 10^{-7} n \frac{1}{K})(K-1)$$

Speedup Model



Speedup
$$^{\bar{p}}(s,n,K) = \frac{T^{\bar{p}}(s,n,1)}{T^{\bar{p}}(s,n,K)}$$



What is the maximum speedup that the program can acheive?

What is the maximum speedup that the program can acheive?

What is the maximum speedup that the program can acheive?

Find the value of K that results in the minimum running time.

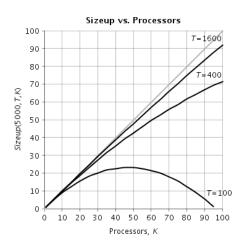
What is the maximum speedup that the program can acheive?

Find the value of K that results in the minimum running time.

Differentiate $T^{\bar{p}}(s, n, K)$ with respect to K, set the derivative equal to 0, and solve for K.

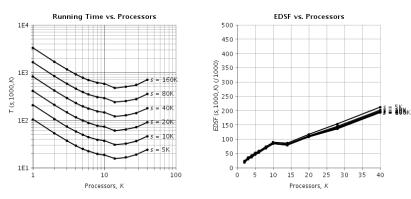
$$\mathcal{K}_{ ext{best}}^{ar{p}}(s,n) = \sqrt{(an^2 - rac{1}{B}128n)/L}$$
 $\mathcal{K}_{ ext{best}}^{ar{p}}(s,n) = \sqrt{9.81 imes 10^{-5}n^2 - 6.59 imes 10^{-4}n}$
 $\mathcal{K}_{ ext{best}}^{ar{p}}(5000, 2000) \approx 20$
 $Speedup^{ar{p}}(5000, 2000, 20) = 9.891$
 $Eff^{ar{p}}(5000.2000, 20) = 0.495$

Sizeup Model



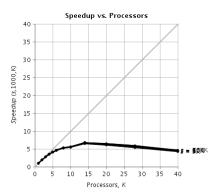
$$Sizeup^{\bar{p}}(s,T,K) = \frac{N^{\bar{p}}(s,T,K)}{N^{\bar{p}}(s,T,1)} = \left(\frac{n^{\bar{p}}(s,T,K)}{n^{\bar{p}}(s,T,1)}\right)^2$$

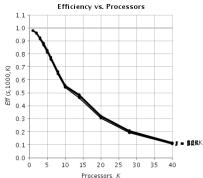
AntiprotonClu Running Time and EDSF



s varying; n = 1000

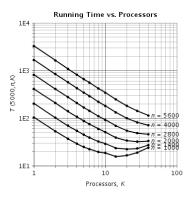
AntiprotonClu Speedup and Efficiency

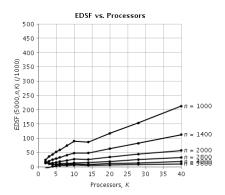




s varying; n = 1000

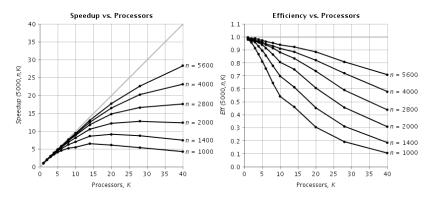
AntiprotonClu Running Time and EDSF





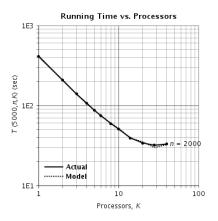
s = 5000; n varying

AntiprotonClu Speedup and Efficiency



s = 5000; n varying

AntiprotonClu Experimental Results



$$s = 5000$$
; $n = 2000$

Parallel computing:

- ▶ bigger problems
- faster answers

Parallel computing:

- bigger problems
- faster answers

Scalability: the *ability* of a parallel program to calculate larger problems as the number of processors *scales up* (increases).

- a qualitative quantity
- ▶ influenced by a number of *quantitative* quantities

Scalability: the ability of a parallel program to calculate larger problems as the number of processors scales up (increases).

- ▶ isoefficiency function: formula for N as a function of K such that the parallel program's efficiency is held the same.
- ightharpoonup size function (N(T, K)): formula for N as a function of K such that the parallel program's running time is held the same.
 - problem size N is proportional to amount of computation (by defn.)

Scalability: the *ability* of a parallel program to calculate larger problems as the number of processors *scales up* (increases).

input-size function (n(T, K)): formula for n as a function K such that the parallel program's running time is held the same.

Scalability: the ability of a parallel program to calculate larger problems as the number of processors scales up (increases).

- input-size function (n(T, K)): formula for n as a function K such that the parallel program's running time is held the same.
 - ightharpoonup problem size N is a function f of input size n
 - ▶ AES key search: $f(n) = 2^n$
 - Floyd's algorithm: $f(n) = n^3$
 - Antiproton motion: $f(n) = n^2$

Scalability: the ability of a parallel program to calculate larger problems as the number of processors scales up (increases).

- input-size function (n(T, K)): formula for n as a function K such that the parallel program's running time is held the same.
 - ightharpoonup problem size N is a function f of input size n
 - ▶ AES key search: $f(n) = 2^n$
 - Floyd's algorithm: $f(n) = n^3$
 - Antiproton motion: $f(n) = n^2$
 - ▶ input size n is a function f^{-1} of problem size N
 - ► AES key search: $f^{-1}(N) = \log_2 N$
 - Floyd's algorithm: $f^{-1}(N) = \sqrt[3]{N}$
 - Antiproton motion: $f^{-1}(N) = \sqrt{N}$

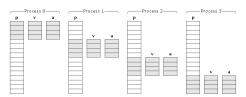
Scalability: the ability of a parallel program to calculate larger problems as the number of processors scales up (increases).

- input-size function (n(T, K)): formula for n as a function K such that the parallel program's running time is held the same.
 - ightharpoonup problem size N is a function f of input size n
 - ▶ AES key search: $f(n) = 2^n$
 - Floyd's algorithm: $f(n) = n^3$
 - Antiproton motion: $f(n) = n^2$
 - input size n is a function f^{-1} of problem size N
 - ► AES key search: $f^{-1}(N) = \log_2 N$
 - Floyd's algorithm: $f^{-1}(N) = \sqrt[3]{N}$
 - Antiproton motion: $f^{-1}(N) = \sqrt{N}$
 - $n(T,K) = f^{-1}(N(T,K))$

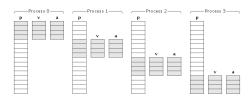
Scalability: the *ability* of a parallel program to calculate larger problems as the number of processors *scales up* (increases).

To solve a given problem size, program requires both *time* and *space*. Often, scalability is limited by the memory required by each process.

Often, scalability is limited by the memory required by each process.



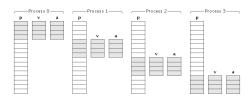
Often, scalability is limited by the memory required by each process.



Memory (in each process) for

- ▶ acceleration array is O(n/K)
- velocity array is O(n/K)
- **•** position array is O(n)

Often, scalability is limited by the memory required by each process.

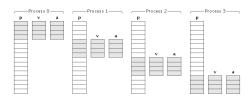


Memory (in each process) for

- lacktriangle acceleration array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- velocity array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- position array is $O(n) = O(\sqrt{K})$

For a fixed running time T and assuming ideal sizeup, n is $O(\sqrt{K})$.

Often, scalability is limited by the memory required by each process.

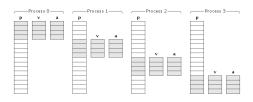


Memory (in each process) for

- lacktriangle acceleration array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- velocity array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- position array is $O(n) = O(\sqrt{K})$

For a fixed running time T and assuming ideal sizeup, n is $O(\sqrt{K})$.

Often, scalability is limited by the memory required by each process.



Memory (in each process) for

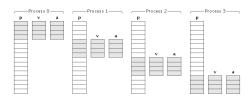
- ▶ acceleration array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- velocity array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- position array is $O(n) = O(\sqrt{K})$

For a fixed running time T and assuming ideal sizeup, n is $O(\sqrt{K})$. In the limit, as $K \to \infty$?

Often, scalability is limited by the memory required by each process.

If K_{best} (for either speedup or sizeup) implies memory required by each process exceeds available memory, then can't acheive best performance.

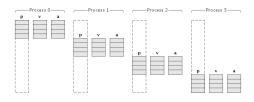
Often, scalability is limited by the memory required by each process.



Memory (in each process) for

- ▶ acceleration array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- velocity array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- position array is $O(n) = O(\sqrt{K}) = O(\sqrt{K})$

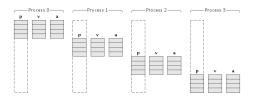
Often, scalability is limited by the memory required by each process.



Memory (in each process) for

- ▶ acceleration array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- velocity array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- ▶ position array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$

Often, scalability is limited by the memory required by each process.



Memory (in each process) for

- ▶ acceleration array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- velocity array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$
- ▶ position array is $O(n/K) = O(\sqrt{K}/K) = O(1/\sqrt{K})$

But, every process needs all positions to compute electrostatic forces.

Rearrange the program's calculation and communication.

Current arrangement:

- ▶ Calculate *all* of the electrostatic forces, then
- ► Commmunicate all of the particle positions (with allGather)

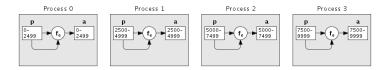
Rearrange the program's calculation and communication.

Current arrangement:

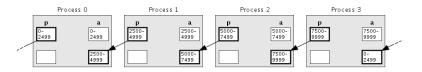
- ▶ Calculate all of the electrostatic forces, then
- ► Commmunicate all of the particle positions (with allGather)

Desired arragement:

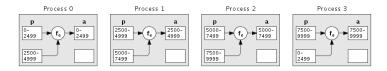
- ► Calculate *a slice* of the electrostatic forces, then
- Communicate a slice of the particle positions, then
- ► Loop



Compute the electrostatic forces between all particles in this process's slice and all particles in this process's slice, acculumating the forces into the acceleration slice.

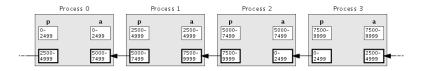


Send this process's slice of positions to the previous process and receive the next process's slice of positions into auxiliary buffer. (There is also another auxiliary buffer.)



Swap the auxiliary buffers.

Compute the electrostatic forces between all particles in this process's slice and all particles in the just-received slice of positions, acculumating the forces into the acceleration slice.

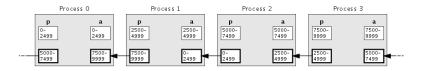


Send the slice of positions in 1^{st} auxiliary buffer to the previous process and receive a slice of positions from the next process into 2^{nd} auxiliary buffer.

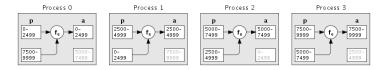


Swap the auxiliary buffers.

Compute the electrostatic forces between all particles in this process's slice and all particles in the just-received slice of positions, acculumating the forces into the acceleration slice.



Send the slice of positions in 1^{st} auxiliary buffer to the previous process and receive a slice of positions from the next process into 2^{nd} auxiliary buffer.



Swap the auxiliary buffers.

Compute the electrostatic forces between all particles in this process's slice and all particles in the just-received slice of positions, acculumating the forces into the acceleration slice.



Swap the auxiliary buffers.

Compute the electrostatic forces between all particles in this process's slice and all particles in the just-received slice of positions, acculumating the forces into the acceleration slice.

Every process must send a slice of positions to one process and receive a slice of positions from another process.

Which collective communication operation?



Swap the auxiliary buffers.

Compute the electrostatic forces between all particles in this process's slice and all particles in the just-received slice of positions, acculumating the forces into the acceleration slice.

Every process must send a slice of positions to one process and receive a slice of positions from another process.

Which collective communication operation? send-receive

```
int toRank = (rank - 1 + size) % size;
int fromRank = (rank + 1) % size;
world.sendReceive (toRank, outBuf, fromRank, inBuf);
```

```
int toRank = (rank - 1 + size) % size;
int fromRank = (rank + 1) % size;
world.sendReceive (toRank, outBuf, fromRank, inBuf);
```

Sends the outgoing message and receives the incoming message simultaneousy in separate threads.

```
int toRank = (rank - 1 + size) % size;
int fromRank = (rank + 1) % size;
world.sendReceive (toRank, outBuf, fromRank, inBuf);
```

Sends the outgoing message and receives the incoming message simultaneousy in separate threads.

$$T_{\text{send-receive}}(b, K) =$$

```
int toRank = (rank - 1 + size) % size;
int fromRank = (rank + 1) % size;
world.sendReceive (toRank, outBuf, fromRank, inBuf);
```

Sends the outgoing message and receives the incoming message simultaneousy in separate threads.

$$T_{\text{send-receive}}(b, K) = (L + \frac{1}{B}b)$$

Note: Assumes same latency and bandwith for both 1 send and K simultaneous sends.

AntiprotonClu2.java

code/AntiprotonClu2.java

AntiprotonClu2.java

code/AntiprotonClu2.java

Before looking at the running-time measurements for AntiprotonClu2, derive a model to predict the running time.

What will we need?

- ▶ a calculation time model (same as AntiprotonClu)
- a communication time model

Calculation Time Model

Antiproton motion algorithm is $O(sn^2)$, where n is the number of particles and s is total number of steps ($s = snaps \cdot steps$).

$$T_{\rm calc}^{\bar{p}2}(s,n,1)=asn^2$$

$$T_{\mathrm{calc}}^{\bar{p}2}(s, n, K) = asn^2 \frac{1}{K}$$

$$T_{\rm calc}^{\bar{p}2}(s,n,1) = 2.04 \times 10^{-8} sn^2$$

$$T_{\rm calc}^{\bar{p}2}(s, n, K) = 2.04 \times 10^{-8} sn^2 \frac{1}{K}$$

- ▶ How many send-receives are performed?
- ▶ How many bits in each send-receives message?

- ▶ How many send-receives are performed? s(K-1)
- ► How many bits in each send-receives message? 128 $n\frac{1}{K}$

- ▶ How many send-receives are performed? s(K-1)
- ► How many bits in each send-receives message? 128 $n_{\overline{k}}^{1}$

$$T_{\text{comm}}^{\bar{p}2}(s, n, K) = s(K - 1)T_{\text{send-receive}}(128n\frac{1}{K}, K)$$

$$T_{\text{comm}}^{\bar{p}2}(s, n, K) = s(K - 1)(L + \frac{1}{B}128n\frac{1}{K})$$

$$T_{\text{comm}}^{\bar{p}2}(s, n, K) = s(K - 1)(2.08 \times 10^{-4} + 1.37 \times 10^{-7}n\frac{1}{K})$$

- ▶ How many send-receives are performed? s(K-1)
- ► How many bits in each send-receives message? 128 $n_{\overline{k}}^{1}$

$$T_{\text{comm}}^{\bar{p}2}(s, n, K) = s(K - 1)T_{\text{send-receive}}(128n\frac{1}{K}, K)$$

$$T_{\text{comm}}^{\bar{p}2}(s, n, K) = s(K - 1)(L + \frac{1}{B}128n\frac{1}{K})$$

$$T_{\text{comm}}^{\bar{p}2}(s, n, K) = s(K - 1)(2.08 \times 10^{-4} + 1.37 \times 10^{-7}n\frac{1}{K})$$

Note: $T_{\text{comm}}^{\bar{p}2}(s, n, K) = T_{\text{comm}}^{\bar{p}}(s, n, K)$.

- ▶ How many send-receives are performed? s(K-1)
- ► How many bits in each send-receives message? 128 $n\frac{1}{K}$

$$T_{\text{comm}}^{\bar{p}2}(s, n, K) = s(K - 1)T_{\text{send-receive}}(128n\frac{1}{K}, K)$$

$$T_{\text{comm}}^{\bar{p}2}(s, n, K) = s(K - 1)(L + \frac{1}{B}128n\frac{1}{K})$$

$$T_{\text{comm}}^{\bar{p}2}(s, n, K) = s(K - 1)(2.08 \times 10^{-4} + 1.37 \times 10^{-7}n\frac{1}{K})$$

Note:
$$T_{\text{comm}}^{\bar{p}2}(s, n, K) = T_{\text{comm}}^{\bar{p}}(s, n, K)$$
.

AntiprotonClu2's send-receive messages are exactly the same as AntiprotonClu's all-gather messages; the former does them interpsersed with calculation, while the later does them all at once.

$$T^{\bar{p}2}(s,n,K) = T_{\mathrm{calc}}^{\bar{p}2}(s,n,K) + T_{\mathrm{comm}}^{\bar{p}2}(s,n,K)$$

$$T^{\bar{p}2}(s, n, K) = asn^2 \frac{1}{K} + s(L + \frac{1}{B}128n\frac{1}{K})(K - 1)$$

$$T^{\bar{p}2}(s,n,K) = 2.04 \times 10^{-8} sn^2 \frac{1}{K} + s(K-1)(2.08 \times 10^{-4} + 1.37 \times 10^{-7} n \frac{1}{K})$$

$$T^{\bar{p}2}(s,n,K) = T_{\mathrm{calc}}^{\bar{p}2}(s,n,K) + T_{\mathrm{comm}}^{\bar{p}2}(s,n,K)$$

$$T^{\bar{p}2}(s, n, K) = asn^2 \frac{1}{K} + s(L + \frac{1}{B}128n\frac{1}{K})(K - 1)$$

$$T^{\bar{p}2}(s,n,K) = 2.04 \times 10^{-8} sn^2 \frac{1}{K} + s(K-1)(2.08 \times 10^{-4} + 1.37 \times 10^{-7} n \frac{1}{K})$$

Note: $T^{\bar{p}2}(s, n, K) = T^{\bar{p}}(s, n, K)$.

Predict that AntiprotonClu2 has exactly the same performance as AntiprotonClu.

$$T^{\bar{p}2}(s,n,K) = T_{\mathrm{calc}}^{\bar{p}2}(s,n,K) + T_{\mathrm{comm}}^{\bar{p}2}(s,n,K)$$

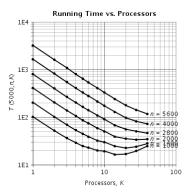
$$T^{\bar{p}2}(s, n, K) = asn^2 \frac{1}{K} + s(L + \frac{1}{B}128n\frac{1}{K})(K - 1)$$

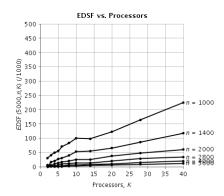
$$T^{\bar{p}2}(s,n,K) = 2.04 \times 10^{-8} sn^2 \frac{1}{K} + s(K-1)(2.08 \times 10^{-4} + 1.37 \times 10^{-7} n \frac{1}{K})$$

Note: $T^{\bar{p}2}(s, n, K) = T^{\bar{p}}(s, n, K)$.

Predict that AntiprotonClu2 has exactly the same performance as AntiprotonClu. But former has significantly improved memory scalability.

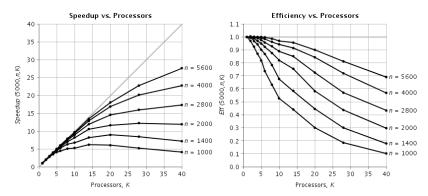
AntiprotonClu2 Running Time and EDSF





s = 5000; n varying

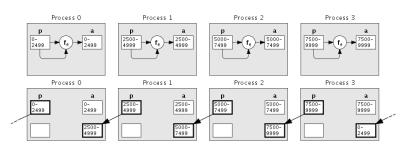
AntiprotonClu2 Speedup and Efficiency



s = 5000; n varying

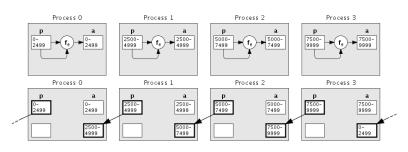
Interspersed Computation and Communication

AntiprotonClu2 improved the memory scalability through pipelined message passing, which interpsersed computation and communication.



Interspersed Computation and Communication

AntiprotonClu2 improved the memory scalability through pipelined message passing, which interpsersed computation and communication.



While communicating, process is not computing!

Further rearrange the program's calculation and communication.

Current arrangement:

- ► Calculate a slice of the electrostatic forces, then
- ► Communicate a slice of the particle positions, then
- ► Loop

Further rearrange the program's calculation and communication.

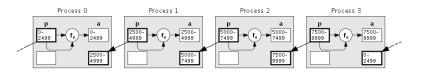
Current arrangement:

- ► Calculate a slice of the electrostatic forces, then
- ▶ Communicate a slice of the particle positions, then
- Loop

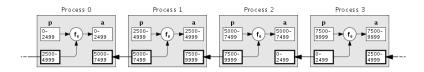
Desired arragement:

- Calculate a slice of the electrostatic forces, and
- Communicate a slice of the particle positions, then
- Loop

(There is also another auxiliary buffer.)



Compute the electrostatic forces between all particles in this process's slice and all particles in this process's slice, acculumating the forces into the acceleration slice; and, at the same time, send this process's slice of positions to the previous process and receive the next process's slice of positions into auxiliary buffer.



Swap the auxiliary buffers.

Compute the electrostatic forces between all particles in this process's slice and all particles in the just-received slice of positions,

acculumating the forces into the acceleration slice;

and, at the same time,

send the slice of positions in 1^{st} auxiliary buffer to the previous process and receive a slice of positions from the next process into 2^{nd} auxiliary buffer.



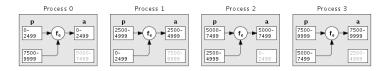
Swap the auxiliary buffers.

Compute the electrostatic forces between all particles in this process's slice and all particles in the just-received slice of positions,

acculumating the forces into the acceleration slice;

and, at the same time,

send the slice of positions in 1^{St} auxiliary buffer to the previous process and receive a slice of positions from the next process into 2^{nd} auxiliary buffer.



Swap the auxiliary buffers.

Compute the electrostatic forces between all particles in this process's slice and all particles in the just-received slice of positions, acculumating the forces into the acceleration slice.

Non-Blocking Send-Receive

```
int toRank = (rank - 1 + size) % size;
int fromRank = (rank + 1) % size;
CommRequest req = new CommRequest();
world.sendReceive (toRank, outBuf, fromRank, inBuf, req);
// perform computation
req.waitForFinish();
```

AntiprotonClu3.java

code/AntiprotonClu3.java

AntiprotonClu3.java

code/AntiprotonClu3.java

Before looking at the running-time measurements for AntiprotonClu3, derive a model to predict the running time.

What will we need?

- ▶ a calculation time model (same as AntiprotonClu)
- ▶ a communication time model (same as AntiprotonClu2)

Calculation Time Model

Antiproton motion algorithm is $O(sn^2)$, where n is the number of particles and s is total number of steps $(s = snaps \cdot steps)$.

$$T_{\mathrm{calc}}^{\bar{p}3}(s,n,1) = asn^2$$

$$T_{\mathrm{calc}}^{\bar{p}3}(s,n,K) = asn^2 \frac{1}{K}$$

$$T_{\rm calc}^{\bar{p}3}(s,n,1) = 2.04 \times 10^{-8} sn^2$$

$$T_{\rm calc}^{\bar{p}3}(s, n, K) = 2.04 \times 10^{-8} sn^2 \frac{1}{K}$$

- ▶ How many send-receives are performed? s(K-1)
- ► How many bits in each send-receives message? 128 $n_{\overline{k}}^{1}$

$$T_{\text{comm}}^{\bar{p}3}(s, n, K) = s(K - 1)T_{\text{send-receive}}(128n\frac{1}{K}, K)$$

$$T_{\text{comm}}^{\bar{p}3}(s, n, K) = s(K - 1)(L + \frac{1}{B}128n\frac{1}{K})$$

$$T_{\text{comm}}^{\bar{p}3}(s, n, K) = s(K - 1)(2.08 \times 10^{-4} + 1.37 \times 10^{-7}n\frac{1}{K})$$

$$T^{\bar{p}3}(s,n,K) = T_{\mathrm{calc}}^{\bar{p}3}(s,n,K) + T_{\mathrm{comm}}^{\bar{p}3}(s,n,K)$$

Running Time Model

$$T^{\bar{p}3}(s,n,K) = \max(T^{\bar{p}3}_{\mathrm{calc}}(s,n,K), T^{\bar{p}3}_{\mathrm{comm}}(s,n,K))$$

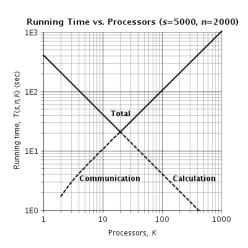
Running Time Model

$$T^{\bar{p}3}(s,n,K) = \max(T^{\bar{p}3}_{\mathrm{calc}}(s,n,K), T^{\bar{p}3}_{\mathrm{comm}}(s,n,K))$$

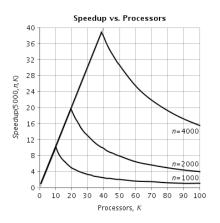
$$T^{\bar{p}3}(s,n,K) = \max(\ asn^2 \frac{1}{K} \ , \ s(K-1)(L+\frac{1}{B}128n\frac{1}{K}) \)$$

$$T^{ar{p}3}(s,n,K) = \max(\ 2.04 \times 10^{-8} sn^2 rac{1}{K} \ , \ s(K-1)(2.08 \times 10^{-4} + 1.37 \times 10^{-7} n rac{1}{K}) \)$$

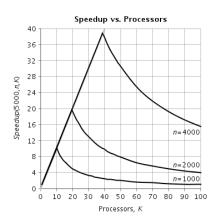
Running Time Model



$$\textit{T}^{\bar{p}3}(s,n,K) = \max(\ 2.04 \times 10^{-8} \textit{sn}^2 \frac{1}{K} \ , \ \textit{s}(K-1)(2.08 \times 10^{-4} + 1.37 \times 10^{-7} n \frac{1}{K}) \)$$



$$Speedup^{\bar{p}3}(s,n,K) = \frac{T^{\bar{p}3}(s,n,1)}{T^{\bar{p}3}(s,n,K)}$$



What is the maximum speedup that the program can acheive?

What is the maximum speedup that the program can acheive?

What is the maximum speedup that the program can acheive?

Find the value of K that results in the minimum running time.

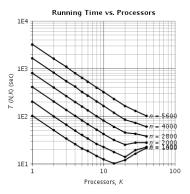
What is the maximum speedup that the program can acheive?

Find the value of K that results in the minimum running time.

Set $T_{\mathrm{calc}}^{\bar{p}3}(s,n,K)$ equal to $T_{\mathrm{comm}}^{\bar{p}3}(s,n,K)$ and solve for K.

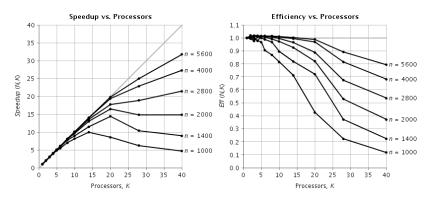
$$\mathcal{K}_{\mathrm{best}}^{ar{p}3}(s,n) = \cdots$$
 $\mathcal{K}_{\mathrm{best}}^{ar{p}3}(s,n) = \cdots$
 $\mathcal{K}_{\mathrm{best}}^{ar{p}3}(5000,2000) \approx 20$
 $Speedup^{ar{p}3}(5000,2000,20) = 20$
 $Eff^{ar{p}3}(5000,2000,20) = 1.000$

AntiprotonClu3 Running Time



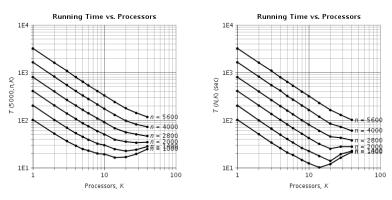
s = 5000; n varying

AntiprotonClu3 Speedup and Efficiency



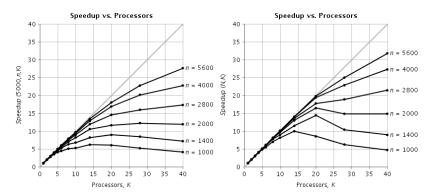
s = 5000; n varying

AntiprotonClu vs. AntiprotonClu3 Running Time



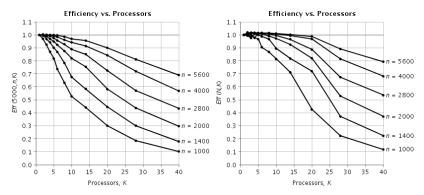
s = 5000; n varying

AntiprotonClu vs. AntiprotonClu3 Speedup



s = 5000; n varying

AntiprotonClu vs. AntiprotonClu3 Efficiency



s = 5000; n varying