# Parallel & Cluster Computing Monte Carlo

Henry Neeman, Director

OU Supercomputing Center for Education & Research University of Oklahoma SC08 Education Program's Workshop on Parallel & Cluster computing August 10-16 2008







#### **Okla. Supercomputing Symposium**



2003 Keynote: Peter Freeman NSF Computer & Information Science & Engineering Assistant Director



2004 Keynote: Sangtae Kim NSF Shared Cyberinfrastructure Division Director



2005 Keynote: Walt Brooks NASA Advanced Supercomputing Division Director

2006 Keynote: Dan Atkins Head of NSF's Office of Cyberinfrastructure



Tue Oct 7 2008 @ OU

Over 250 registrations already!

Over 150 in the first day, over 200 in the first week, over 225 in the first month.

2007 Keynote: Jay Boisseau Director Texas Advanced Computing Center U. Texas Austin



2008 Keynote: José Munoz Deputy Office Director/ Senior Scientific Advisor Office of Cyberinfrastructure National Science Foundation



2

FREE! Parallel Computing Workshop Mon Oct 6 @ OU sponsored by SC08 FREE! Symposium Tue Oct 7 @ OU

http://symposium2008.oscer.ou.edu/



### **Embarrassingly Parallel**

An application is known as *embarrassingly parallel* if its parallel implementation:

- 1. can straightforwardly be broken up into roughly equal amounts of work per processor, **AND**
- 2. has minimal parallel overhead (e.g., communication among processors).

We <u>love</u> embarrassingly parallel applications, because they get <u>near-perfect parallel speedup</u>, sometimes with modest programming effort.

Embarrassingly parallel applications are also known as *loosely coupled*.





#### **Monte Carlo Methods**

Monte Carlo is a European city where people gamble; that is, they play games of chance, which involve randomness.

<u>Monte Carlo methods</u> are ways of simulating (or otherwise calculating) physical phenomena based on randomness.

Monte Carlo simulations typically are embarrassingly parallel.





### Monte Carlo Methods: Example

Suppose you have some physical phenomenon. For example, consider High Energy Physics, in which we bang tiny particles together at incredibly high speeds.

We want to know, say, the average properties of this phenomenon.

There are infinitely many ways that two particles can be banged together.

So, we can't possibly simulate all of them.





#### Monte Carlo Methods: Example

Suppose you have some physical phenomenon. For example, consider High Energy Physics, in which we bang tiny particles together at incredibly high speeds.



There are infinitely many ways that two particles can be banged together.

So, we can't possibly simulate all of them.

**Instead**, we can **randomly choose a finite subset** of these infinitely many ways and simulate only the subset.





### Monte Carlo Methods: Example

Suppose you have some physical phenomenon. For example, consider High Energy Physics, in which we bang tiny particles together at incredibly high speeds.



- There are infinitely many ways that two particles can be banged together.
- We randomly choose a finite subset of these infinitely many ways and simulate only the subset.
- The average of this subset will be close to the actual average.





#### **Monte Carlo Methods**

In a Monte Carlo method, you randomly generate a large number of example cases (*realizations*) of a phenomenon, and then take the average of the properties of these realizations.

When the realizations' average converges (i.e., doesn't change substantially if new realizations are generated), then the Monte Carlo simulation stops.





## **MC: Embarrassingly Parallel**

Monte Carlo simulations are embarrassingly parallel, because each realization is completely independent of all of the other realizations.

That is, if you' re going to run a million realizations, then:

- 1. you can straightforwardly break up into roughly  $1M / N_p$  chunks of realizations, one chunk for each of the  $N_p$  processors, **AND**
- 2. the only parallel overhead (e.g., communication) comes from tracking the average properties, which doesn't have to happen very often.





#### **Serial Monte Carlo**

```
Suppose you have an existing serial Monte Carlo simulation:
PROGRAM monte_carlo
CALL read_input(...)
DO realization = 1, number_of_realizations
CALL generate_random_realization(...)
CALL calculate_properties(...)
END DO
CALL calculate_average(...)
END PROGRAM monte_carlo
How would you parallelize this?
```





#### **Parallel Monte Carlo**

```
PROGRAM monte carlo
 [MPI startup]
  IF (my rank == server rank)
                                 THEN
    CALL read input(...)
  END IF
  CALL MPI Bcast(...)
  DO realization = 1, number of realizations
    CALL generate random realization(...)
    CALL calculate realization properties (...)
    CALL calculate local running average(...)
  END DO
  IF (my rank == server rank)
                                 THEN
     [collect properties]
  ELSE
     [send properties]
  END IF
  CALL calculate global average from local averages (...)
  CALL output overall average(...)
 [MPI shutdown]
END PROGRAM monte carlo
```





#### **Okla. Supercomputing Symposium**



2003 Keynote: Peter Freeman NSF Computer & Information Science & Engineering Assistant Director



2004 Keynote: Sangtae Kim NSF Shared Cyberinfrastructure Division Director



2005 Keynote: Walt Brooks NASA Advanced Supercomputing Division Director

2006 Keynote: Dan Atkins Head of NSF' s Office of Cyberinfrastructure



Tue Oct 7 2008 @ OU

Over 250 registrations already!

Over 150 in the first day, over 200 in the first week, over 225 in the first month.

2007 Keynote: Jay Boisseau Director Texas Advanced Computing Center U. Texas Austin



2008 Keynote: José Munoz Deputy Office Director/ Senior Scientific Advisor Office of Cyberinfrastructure National Science Foundation



#### FREE! Parallel Computing Workshop Mon Oct 6 @ OU sponsored by SC08 FREE! Symposium Tue Oct 7 @ OU

http://symposium2008.oscer.ou.edu/



#### **To Learn More**

#### <u>http://www.oscer.ou.edu/</u> <u>http://www.sc-conference.org/</u>







# Thanks for your attention!

**Questions?** 

