# Projections: Scalable Performance Analysis and Visualization

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 $\rightarrow$  Charm++

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- Runtime system schedules which method to execute next (based on messages that have arrived)



 $\rightarrow$  Collections of Objects

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#### Charm++

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- Often communication patterns can be represented nicely by interactions between a collection of elements
- Objects can be organized into typed, indexed collections
  - Dense
  - Sparse
  - Multi-dimensional (1d-6d)
  - Elements can be dynamically inserted into or deleted

#### Charm++

#### $\rightarrow$ Collections of Objects





- Many more objects than processors
  - Anywhere from tens to hundreds per processor
- Fine-grained resolution of events
  - May be as small as tens of microseconds per event
- Logical entities (objects) are distinct from physical (processors)
  - Mapping may change over time



- Most of the code is written in C++
- Parallel objects have a corresponding parallel interface in a .ci file
- The .ci file is translated to C++ code
  - We have some compiler level support we can leverage

# Methodology

→ Event Tracing

#### Trace-based instrumentation of events

- Certain methods in the system are marked as *entry* methods
  - ★ Meaning they can be invoked remotely
  - \* These remote methods are automatically traced by the system
- Messages sent and received
- System events
  - Certain scheduler-level events or system states are recorded: processor idleness, communication overhead, message serialization, etc.

# **User Intervention**

 $\rightarrow$  Event Tracing

#### Language gives flexibility to the user

- Methods can be annotated by the notrace attribute, which causes the code generation to eliminate tracing overhead altogether
- Non-entry methods (not traced by default), can be annotated as local to automatically add tracing
- API provides further control to the programmer
  - Turn tracing on or off
    - ★ On a subset of the processors or objects
    - ★ During some times
  - Register user-defined functions for tracing
  - Trace point events or bracketed events (register name and then call API when it occurs)
  - Save memory usage at a point in the program execution

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## **Charm++: Runtime Data Collection**

- Charm++ has several strategies built-in that have varying data/memory overheads
  - Full tracing
    - An event is composed of the time, sending/receiving processor, entry method, object, etc.
    - Each event is logged per processor in memory and then is incrementally written to disk
  - Summary
    - Each processor is allotted a fixed number of equally sized time bins that hold averages over the time range



- Research on this began in 1992
- Java-based visualization tool that reads traces (summary or full)
- Supports many different ways of visualizing the data
- Scaling
  - Tested with over 100k cores
  - It is multi-threaded and has been optimized for memory usage
- How to use it
  - Download the .jar, works out of the box with Charm++
  - Link with the flag -tracemode projections
  - sit://charm.cs.uiuc.edu/projections.git
- Support beyond Charm++
  - We are actively improving the prototyped MPI tracing layer
  - Support for Global Arrays exists in alpha form

#### **Timeline**



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### Timeline

 $\rightarrow$  NAMD: Apoa1 system, 92k atoms, 32k cores, about 3 atoms per core!



#### **Time Profile**

ightarrow NAMD: Apoa1 system, 92k atoms, no communication thread



#### **Time Profile**

ightarrow NAMD: Apoa1 system, 92k atoms, with communication thread



#### Histogram

 $\rightarrow$  NAMD: Apoa1 system, 92k atoms, 1-away decomposition



### Histogram

 $\rightarrow$  NAMD: Apoa1 system, 92k atoms, 2-away decomposition



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#### **Time Profile**

ightarrow NAMD: Apoa1 system, 92k atoms, with communication thread



## **Usage Profile**



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## **Communication Over Time**



### **Outlier/Extrema View**



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### **Timeline**

#### $\rightarrow~$ Colored by memory for LU



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### **Profile Memory Scatter**



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## **Live Analysis**

#### Can we monitor performance as the application is actually running?

- Uses the Converse client/Server interface
  - \* We can interact with the runtime as the program runs using python
  - Allows us to stream performance data to Projections
- Demo: utilization

## **End-of-run Analysis**

- When we scale over 100k cores the data becomes very large and unmanageable
- Deathbed analysis
  - Use the full parallel machine at the end of the execution for some analysis
  - e.g. k-means clustering to pick out exemplar processors
- We are currently developing algorithms for this

## Conclusion

- Projections
  - We are constantly improving it
  - A mature tool that grew over the years out of necessity
- We are not experts in graphics or visualization
  - As the number of cores increases along with data volume, we need better techniques and help from the broader community