Intuitive Visualizations for Performance Analysis at Scale

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Massive parallelism has made performance a data-rich field, but we lack the tools to interpret and understand the data

- We can collect data at scale, but data volume is overwhelming
 - Too many variables to measure
 - Difficult to write out data from 500 million cores, even if we do measure it
- Information is highly categorical, discontinuous
 - Profiles, traces
 - Hardware Performance Counters
 - FP counts, cache misses, network traffic
 - Counts map to particular cores
- MPI Process ID space is often unintuitive
 - Rank offers little insight into underlying network
- It is difficult to apply analysis techniques because this data lacks structure



MPI Trace Data from runs with 16 and 34 processes



Floating Point Instruction Counts



Presenting performance data on more familiar, intuitive domains provides opportunities for new insights

- How can we make raw performance data more useful?
- Application developers understand application data
 - Temperature plot of Miranda data
- What if we could map performance attributes onto application data structures?
- We constructed a simple example to show FP data in the application domain
 - Can show cache misses similarly
- This is only the tip of the iceberg
 - We must extend this approach to more domains and more complex data
 - We will enable feature-based analysis and correlations





PAVE will develop methods to organize and analyze performance data in domains familiar to scientists

PAVE will develop new analysis techniques to:

- Attribute performance measurements to intuitive domains: *Physical, Hardware*, and *Communication*
- Extract features within data domains
- Correlate features between domains
- Analyze mappings among domains
- This work can only be accomplished by combining and extending state-of-the-art techniques from two fields:
 - We will extend run-time performance analysis for application-semantic attribution at scale
 - We will restructure performance data so that it is amenable to analysis
 - We will develop analysis techniques to correlate and map features between these domains
- Data domains, the correlations between them, and the analysis of their mappings will provide new insights into application performance





PAVE Overview

1. Hardware to Application mapping

Data-dependent computation in fluid dynamics simulations

2. Communication Visualization for AMR

• Visualizing bottlenecks in

3. Boxfish network visualization tool

• Plotting network counters on the network

4. Future Directions

- Higher resolution hardware to application mapping
- Adding structure back to parallel trace visualizations



Projections on the app domain



Aluminum distribution



Velocity distribution



L1 cache misses



Mystery of L1 misses

- One core on each socket has more L1 misses
- Caused by execution of MPI collective operations
- Need for different perspectives to disambiguate causes





We are developing techniques to automatically segment features in performance data



L1 Cache Misses

Same data with linear color map



FP Operations



We are developing techniques to automatically segment features in performance data

Same data with linear color map



L1 Cache Misses with MPI worker filtered



FP Operations



We are developing techniques to automatically segment features in performance data

Same data with linear color map



L1 Cache Misses with MPI worker filtered



FP Operations



L1 Misses per FP operation: Proxy for efficiency



Projections on the comm domain

- Case study: SAMRAI, structured adaptive mesh refinement
 - Blue Gene/P at Argonne (256 to 128K cores)





Timings in MPI rank spaceBottleneck is in phase 1 and not phase 3





We plotted performance metrics on the SAMRAI communication graph

Load (Cells per process) Before balancing

Time spent redistributing boxes



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1024 processes

We developed a scalable visualization for large communication graphs



Load on 16k cores

Wait time for box distribution

Wait time with flow information

This shows data for 16,384 cores



Our visualization can be adaptively refined for more detail

- Angles are apportioned by flow in the subtree
- Heavier trees are expanded to a deeper level
- Can see flow problems at any level of the tree

1024-process expanded tree



Performance improvements

- Mitigate the problem by reducing the size of box metadata
- Trade off slightly increased imbalance for coarser boxes
- Leads to 50% reduction in load balancing time at 64k cores
 - 22% reduction in overall time on 65k cores



We anticipate that network topology will become a performance issue once load balance is O(log(P))

 We have mapped network measurements to SAMRAI patches

 Plots show patches colored by maximum hops on the physical network to any neighbor patch

 SAMRAI LinAdv benchmark does not appear to be affected by this imbalance, but other codes will be.





We have developed the Boxfish visualization tool to better understand network performance





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Overview Minimaps for giving context once user zooms in. Color based on mean link values for the links in that plane. Three different orientation options as user can look into the 3D view in the X, Y or Z directions. These views also give an overview of the communication behavior

Interactive legend for selecting range of values to be viewed. Axes showing current orientation of the view. Clicking on any of the directions shows links in only that direction.



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Grid layout to highlight planes in the 2D view with mouse over. The 2D view also supports interactions like zooming and translations



The 2D view can display all nodes without any occlusion. Only half of X links and half of Y links and all of Z links are shown. The Z links are along the diagonals.



3D view supports interactions like zooming, rotation and translation.



We can use Boxfish to visualize job layout on Cray machines





Boxfish's 2D mini-maps summarize bandwidth and help to explain performance differences





Future Directions: Fine-grained Application Mapping using PEBS

1000x1000 Matrix Multiply with different Blocking Optimizations





Future Directions: Fine-grained Application Mapping using PEBS Counters

LULESH





Future Directions: Adding Structure Back to Parallel Trace Visualization



Messy Multigrid Trace

• With Real Time



Same trace!

- With logical time steps
- Colors show lateness



