



Working Group Outbrief



Visualization and Data Analysis

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Working Group Description



- **The scope of our working group is scientific visualization and data analysis.**
 - **Scientific visualization**
 - refers to the process of transforming scientific simulation and experimental data into images to facilitate visual understanding
 - **Data analysis**
 - refers to the process of transforming data into an information-rich form via mathematical or computational algorithms to promote better understanding
 - **Data management - shared with IONS**
 - refers to the process of tracking, organizing and enhancing the use of scientific data
- **The purpose of our work is to enable scientific discovery and understanding.**
 - **Our scope includes an exascale software and hardware infrastructure that effectively supports visualization and data analysis.**



Identify current state-of-the-art



- **Production visualization tools scalably process large data**
 - **Suite of data-parallel visualization, analysis and rendering algorithms**
 - Open-source tools – ParaView, Visit
 - Commercial tool – Ensignt
- **ASC success story**
 - **NNSA/ASC created the large-data scientific visualization tools in use by other agencies (NSF, DOD) and around the world**



Identify Exascale Visualization and Data Analysis Needs



- **Visualization and Data Analysis (VDA)**
- **Broad range of scope for VDA**
 - **VDA as an application**
 - **VDA as a service**
 - **VDA as a systems infrastructure**
- **Note: like apps, VDA capabilities will require development to exploit opportunities in evolving platforms**



1. Exascale Challenges – storing a full-range of results for later analysis becomes impossible due to technology trends



- The rate of performance improvement of rotating storage is not keeping pace with compute.
- Provisioning additional disks is a possible mitigation strategy, however, power, cost and reliability issues will become a significant issue.
- A new in-situ exascale visualization and data analysis approach is needed:
 - Slow output to data hierarchy / Data movement = power/cost
 - Where will we process data?
 - Different customer-driven approaches require integration at different HW 'levels'



2. Exascale Challenges - Exascale simulation results must be distilled with quantifiable data reduction techniques



- **Exascale as massive data**
 - **Defacto data reduction technique**
 - ① Visualization algorithms
 - ② Rendering massive numbers of polygons
 - **This puts lots of data into a single pixel, combined by the renderer**
 - This is a workable method but is it what the user wants?
 - **This approach provides the foundation for our current successes**
 - brute force approach that requires significant computing resources
 - difficult to quantify the bias of this approach
- **Approaches that quantifiably reduce data as it is generated need to be explored**



3. Exascale Challenges - New exascale-enabled physics approaches require corresponding new visualization and data analysis approaches



- **Implication of exascale as massive compute**
 - **Statistical physics approaches**
 - Statistical modeling of a physical process
 - **Parametric studies**
 - record how a simulation responds in a parameter space of possibilities
 - **Multi-physics approaches**
 - simulate a linked model of different related phenomena such as a linked physics and chemistry simulation
 - **Multi-scale approaches**
 - simulate phenomena at different spatial and temporal scales
- **Understanding and presenting both summarized and highlighted results from multiple sources is an important technical challenge that needs to be addressed.**



4. Exascale Challenges - Visualization and data analysis approaches will need to run efficiently on exascale platform architectures



- Need to take advantage of a very high degree of parallelism
- Technical challenges include achieving portability, efficiency and integration flexibility with simulation codes



1. Path Forward - New visualization and data analysis software infrastructure



- When?: Run-time, Postprocessing
- How?: Interactive, Batch
- **In-situ analysis within the simulation code**
 - Run-time, (batch or interactive)
- Post-processing --- advanced query-based approach
- **Revolutionary approach**
 - **Phase 1**
 - Prototype approaches in applications
 - **Phase 2&3**
 - Develop and deploy
 - Continue R&D
- **Required Partnership**
 - IONS, tools, systems
 - Apps for co-design
- **Metric**
 - Our success will be measured by our readiness for applications as machine delivery milestones are met
- **Risks**
 - How to do discovery science in an in-situ world?
 - Won't find an effective analysis approach for exascale applications



2. Advanced quantifiable data reduction algorithms



- **Data triage**
 - **How do we significantly reduce the data as it is generated?**
 - Statistical sampling
 - Compression
 - Multi-resolution
 - Science-based feature extraction
- **Revolutionary approach**
 - **Phase 1**
 - Prototype approaches independently and with applications
 - **Phase 2&3**
 - Develop and deploy
 - Continue R&D
- **Required Partnership**
 - Applied math
 - Apps for co-design
- **Metric**
 - **Measure of amount of data reduced and quality of result, time**
- **Risks**
 - **Won't find an effective analysis approach for exascale applications**



3. Visualization and data analysis techniques to help understand advanced exascale physics



- Visualization and Data Analysis for:
 - Statistical physics approaches
 - Parametric studies
 - Multi-physics approaches
 - Multi-scale approaches
- how results from different aspects of a simulation suite relate to each other
- Evolutionary/Revolutionary approach
 - Phase 1
 - Prototype approaches independently and with applications
 - Phase 2&3
 - Develop and deploy
 - Continue R&D
- Required Partnership
 - Applications for co-design
- Metric
 - Ties to appropriate application milestones
- Risks
 - Won't understand output of exascale applications



4. Implement core visualization and data-analysis capability using a scalable parallel infrastructure



- – Our visualization and data analysis solutions need to work on the exascale supercomputers on both swim lanes.
- Evolutionary approach
 - Phase 1
 - Prototype approaches independently and with applications
 - Phase 2&3
 - Develop and deploy
 - Continue R&D
- Required Partnership
 - Programming models, tools and applications groups
- Metric
 - Our success will be measured by our readiness for applications as machine delivery milestones are met
- Risks
 - Not running on the machines



5. Exascale visualization and data analysis hardware infrastructure



- **Data-intensive hardware infrastructure for the exascale platform**
 - Memory buffers for staged analysis and storage
 - Analysis-enabled storage
 - Large node memory portion of the supercomputer
- **Evolutionary/Revolutionary approach**
 - Phase 1
 - Prototype approaches independently and with applications
 - Phase 2&3
 - Develop and deploy
 - Continue R&D
- **Required Partnership**
 - HW, Systems, I/O
- **Metric**
 - Ties to appropriate machine milestones
- **Risks**
 - HW platform that makes data analysis difficult



6. Tracking and using knowledge about the scientific goals makes visualization and data analysis more effective



- **Provenance**
 - Where the data comes from?
 - How was it generated?
 - Uncertainty quantification
 - Workflow management and beyond
- **Monitoring, debugging**
 - Supercomputer “situational awareness”
- **Revolutionary approach**
 - **Phase 1**
 - Prototype approaches independently and with applications
 - **Phase 2&3**
 - Develop and deploy
 - Continue R&D
- **Required Partnership**
 - Tools, Systems, Applications for co-design
- **Metric**
 - Time to solution, cost, improved quality
- **Risks**
 - Won’t know where data came from
 - Inefficient use of supercomputer



Recommended Co-Design Strategy



- **Critical steps/activities**
 - **Data Analysis and Movement are critical cross-cutting issues**
 - **Develop well-defined HW needs for VDA**
 - **In-situ and interactive data extraction could benefit from co-design**
 - (e.g. resource management, time series analysis)
 - **Pilot co-design projects to help develop teams and define co-design processes**
 - **These can be subsets – VDA and IO working with a defined app**
- **Working with vendors**
 - **Well-defined communication with partners and vendors**
 - **Visualization software vendors (e.g. Kitware, CEI), platform vendors**
 - **Path Forward investments**



Recommended Co-Design Strategy



- **Role of skeleton/compact apps**
 - **VDA mini-apps**
 - Investigate coupling of in-situ capabilities with applications at scale
 - **Inform Architecture decisions by providing information on use patterns**
- **Concerns/suggestions**
 - **Organization effects outcomes**
 - Partnering with Verification/Validation, applied math
 - Overlap with other groups
 - **Influence we can have on HW vendors w/in timeframe**
 - Particularly with respect to data
 - **Co-design should have significant investment and sustained commitment**



Big Picture Issues



- **Coordination**
 - VDA will be more tightly coupled with apps than in the past
 - Data movement and provenance are shared responsibilities
- **Test beds**
 - Common test beds crucial to creating Exascale community
 - Specialized test beds may be necessary
- **Simulators**
 - Requirement: include adequate characterization of data movement
 - Unclear if there are requirements for VDA-only simulation components
 - Requirement: system-level simulation, to model various VDA use cases
- **Remaining gaps**
 - Interactive data analysis (e.g. querying and extraction) for Exascale is coupled with in-situ (what you can compute at runtime) and computation on data (what you want to understand after the simulation)



Conclusions



A data-oriented perspective is critical to exascale success



End

