Experiences in Translational Computer Science Research: Data-Management for Extreme Science

Manish Parashar*

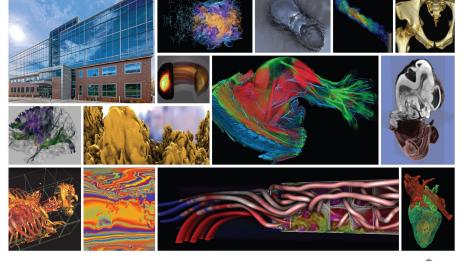
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Presidential Professor, School of Computing University of Utah

PPAM 2022, Gdansk, Poland September 14, 2022











Outline

- Introducing Translational Computer Science (TCS)
- Experiences in TCS Research: Data-Management for Extreme Science
- Important Issues for TCS
- Conclusion

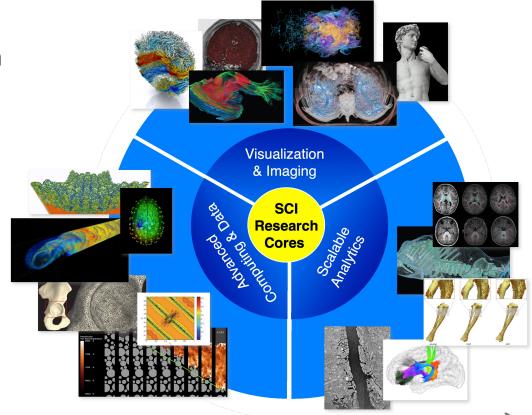




Scientific Computing & Imaging (SCI) Institute

Goal: Transformation of science and society through translational research and innovation in computer, computational and data science

- Multidisciplinary, convergent, collaborative
- Simulation, imaging, visualization, data management/analytics, advanced computing
- Software/system development and distribution

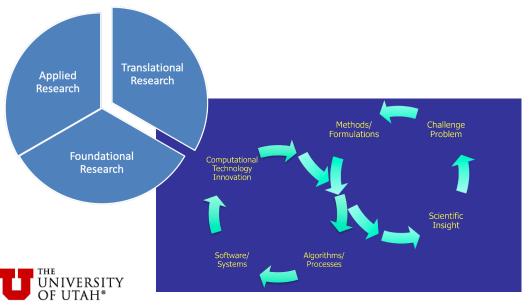


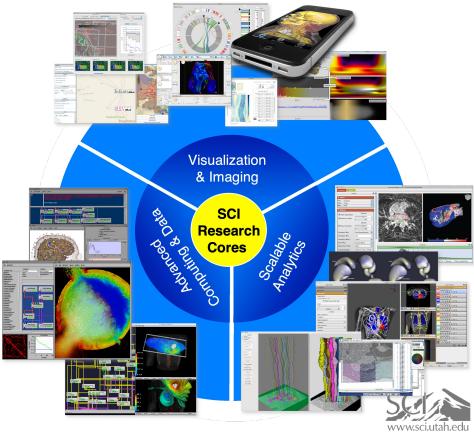




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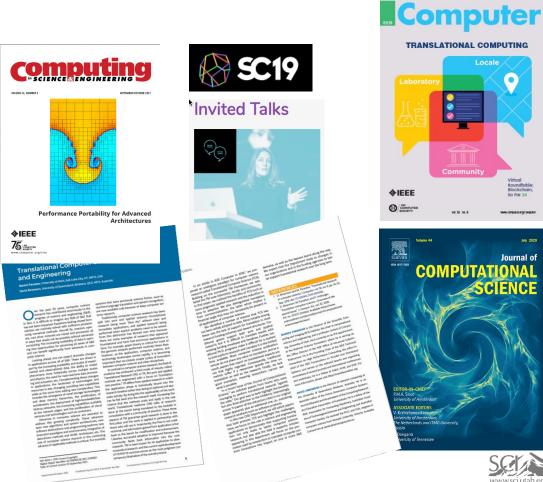


The role of a French chateaux and red wine ... with David Abramson



Workshop on CCDSC, Lyon France, 2018 Jack Dongarra, Bernard Tourancheau









- An interdisciplinary branch of the biomedical field supported by three main pillars:
 - Benchside, Bedside and Community
 - Combines disciplines, resources, expertise, and techniques within these pillars to promote enhancements in prevention, diagnosis, and therapies
- Differs subtly from applied biomedical research, in which a research problem has a potential real-world application
 - Findings are applied as a specific phase of the research plan
 - This not only demonstrates applicability and practicality, but also generates tangible outcomes
- Now well understood and has become a de-facto standard for much of biomedical research
- Intrinsically helps generate outcomes because the research is applied as part of the original plan, as opposed to being an afterthought once the project has completed



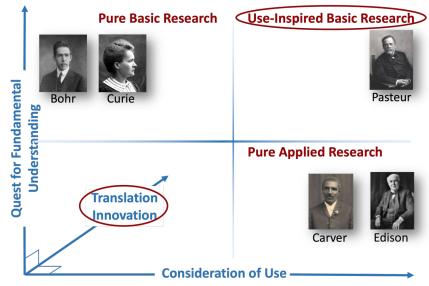


Translational Computer Science (TCS)

TCS refers to research that bridges foundational and use-inspired (and applied) research with the *delivery and deployment of its outcomes* to the target community and supports essential *bi-direction interplays* where delivery and deployment process informs the research

- Motivated by the growing importance of computing and data across all of science and society
- Aimed at accelerating the impact of computer, computational and data science
- Inspired by the definition and impact of Translational Medicine
- Focused on taking research from the Laboratory to the Locale to the Community
 - Laboratory, Locale might be physical or virtual
 - Community: Users and early adopters who work with the technology, and can include public bodies that would help in the evaluation



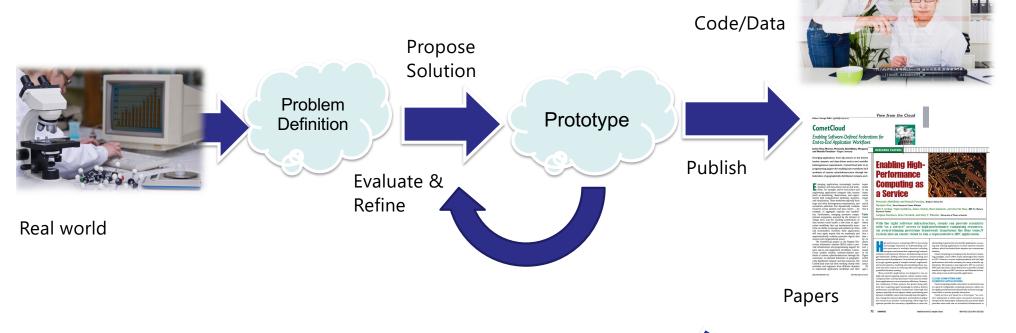


So how does this differ from traditional research pipeline?





Typical CS research workflow

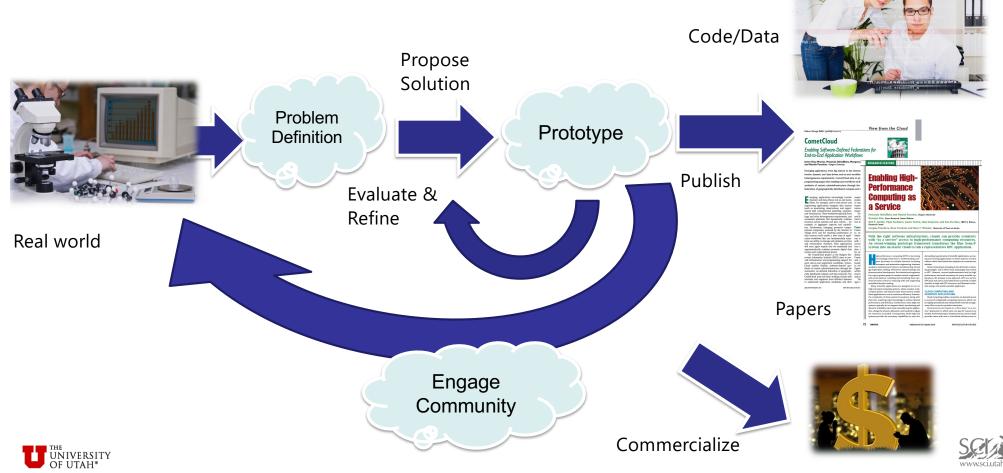








Translational CS research workflow



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Science transformed by extreme scales, pervasive compute, data

- Exascale, and beyond
- Disruptive innovations => new opportunities
- Novel paradigms / a computing continuum
- Unconventional software stacks



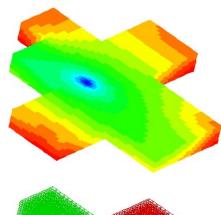


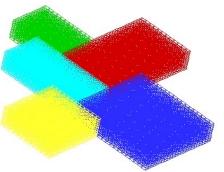
"A supercomputer is a device for turning compute-bound problems into I/O-bound problems" – Ken E. Batcher

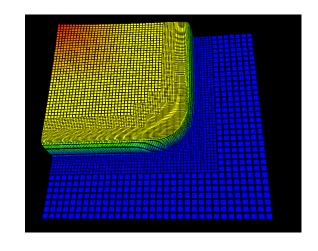


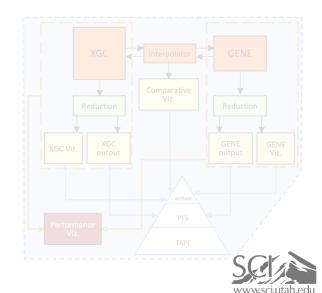
Data Management for Extreme Science: Part I – An Accidental Translationist

- Dynamic, adaptive formulations
- Coupled models and codes
- In-situ workflows/in-transit processing
- End-to-end workflows



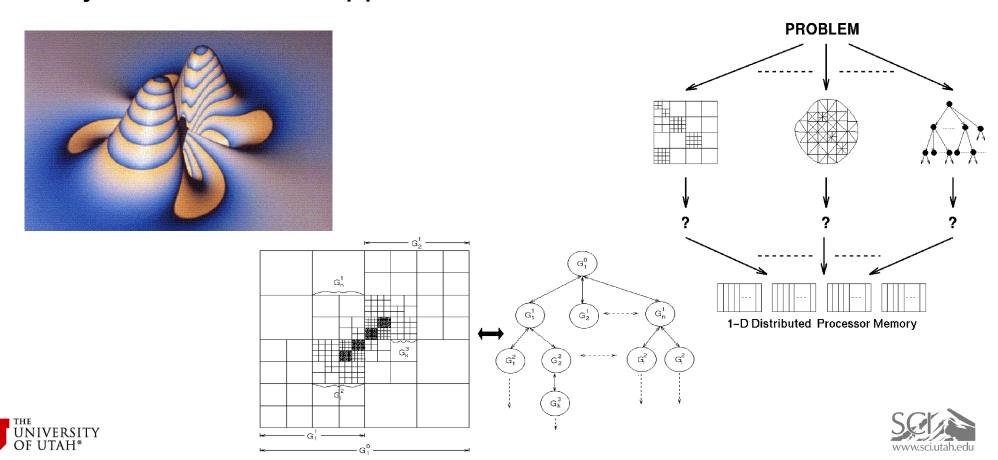








Binary Black Holes, Structured Adaptive Mesh Refinement, & Dynamic Parallel Applications



Data Management for Dynamically Adaptive Applications

- Semantically Specialized DSM
 - Regular access semantics to dynamic, distributed data
 - Encapsulates distribution, communications
- Hierarchical Distributed Dynamic Array / Distributed Shared Objects
 - Hierarchical Index-Space + Extendible Hashing
- Adaptive Run-time Management
 - Application and system sensitive management
 - Policy-based adaptations
- Software Deployed:
 - Grace/Dagh, Mace, Armada



SYSTEMS ENGINEERING FOR HIGH PERFORMANCE COMPUTING SOFTWARE: THE HDDA/DAGH INFRASTRUCTURE FOR IMPLEMENTATION OF PARALLEL STRUCTURED ADAPTIVE MESH

MANISH PARASHAR* AND JAMES C. BROWNE!

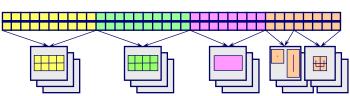
Abstract. This paper defines, describes and illustrates a systems engineering process for development of software systems implementing high performance computing applications. The example which drives the creation of this process is development of a flexible and extendible program development infrastructure for parallel structured adaptive meshes, the HDDA/DAGH package. The fundamental systems engineering principles used (hierarchical abstractions based on separation of concerns) are well-known but are not commonly applied in the context of high performance computing software. Application of these principles will be seen to enable implementation of an infrastructure which combines breadth of applicability and portability with high performance.

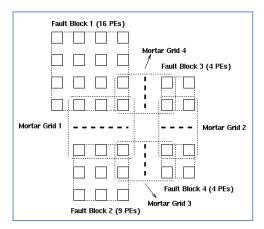


Application Locality



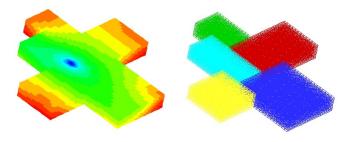
Storage Locality



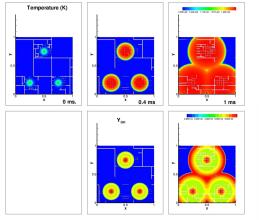




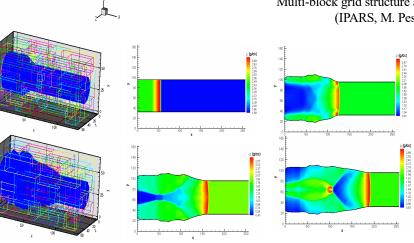
Translations Impacts



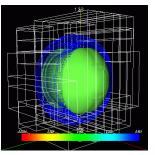
Multi-block grid structure and oil concentrations contours (IPARS, M. Peszynska, UT Austin)

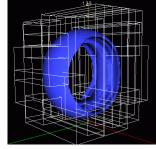


Mixture of H2 and Air in stoichiometric proportions with a non-uniform temperature field (GrACE + CCA, Jaideep Ray, SNL, Livermore)



Richtmyer-Meshkov - detonation in a deforming tube - 3 levels. Z=0 plane visualized on the right (VTF + GrACE, R. Samtaney, CIT)





Blast wave in the presence of a uniform magnetic field) – 3 levels of refinement. (Zeus + GrACE + Cactus, P. Li, NCSA, UCSD)



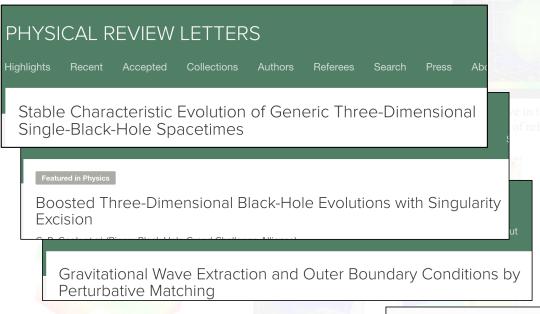
Richtmyer Meshkov (3D) R. Samtaney, CIT



1024x128x128, 3 levels, 2K PE's Time: ~ 15% Memory: ~25%



Translational Impacts



Journal of Physics: Conference Series

OPEN ACCESS

Toward a first-principles integrated simulation of tokamak edge plasmas



Asynchronous replica exchange for molecular simulations

Journal of Physics: Condensed Matter

PAPER

Nucleosome positioning and composition modulate *in silico* chromatin flexibility

SC97·Technical Paper

TOC Authors Sessions Abstracts PostScript

A Common Data Management Infrastructure for Adaptive Algorithms for PDE Solutions

UNIVERSITY OF UTAH*

CIT

Towards Dynamic Data-Driven Management of the Ruby Gulch Waste Repository

SPE 37979

A New Generation EOS Compositional Reservoir Simulator: Part I – Formulation and Discretization.

Variable Resolution Topographic Mapping of Ancient Fluvial Landscapes in Australia

Lost in translation... What happened to DAGH/GrACE, MACE?

- Lost in translation...could not be sustained
 - Translation not part of the research plan
 - Time, effort not allocated for user support, addressing user feedback, new requirements, etc.
 - Lack of funding/support for translation
 - No funding to hire developers/programmers
 - · PhD student theses research does not focus on translational work
 - Academic career-path required reprioritization
 - Academic structures don't value translation

But

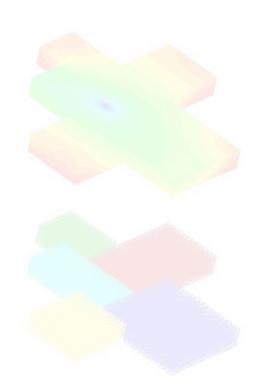
- Effective, had rapid uptake, and impact
- Conceptual framework (and some code) persists, used by other frameworks and influences codebases
- Many lessons learnt that informed subsequent work

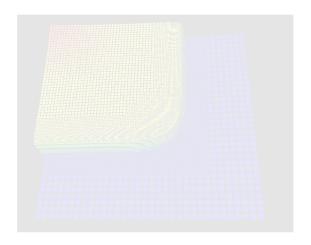


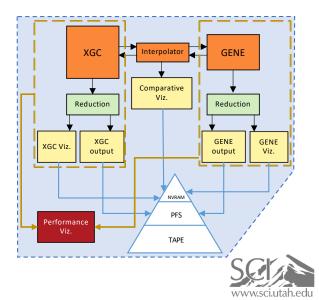


Data management for Extreme Science: Part II

- Dynamic, adaptive formulations
- Coupled models and codes
- In-situ workflows/in-transit processing
- End-to-end workflows

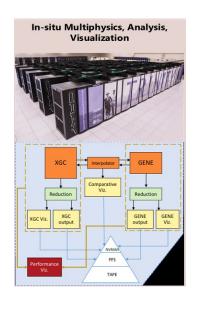


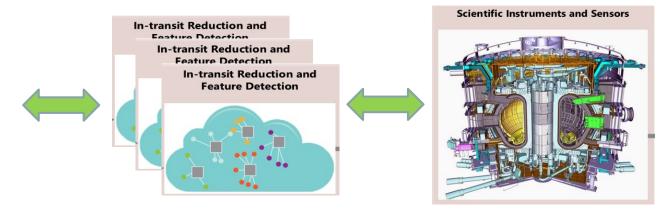






Extreme-Scale, End-to-End Science Workflows





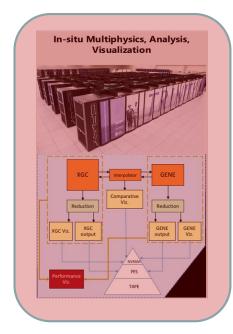
Plasma Disruption Analysis Fusion Workflow

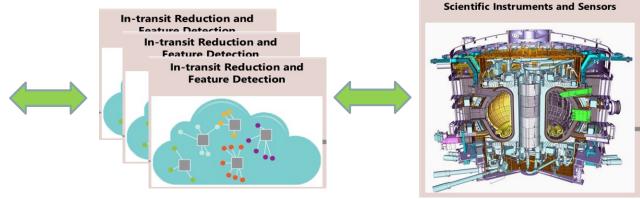
- The goal is to implement a plasma visualization diagnostics system that enables early prediction of anomalies while the tokamak operates
- Data is a time-series of matrix images obtained via direct 2D/3D visualizations of the tokamak plasma
- Analytics include structural analysis and blob detection, filtering & visualization





Rethinking Extreme-Scale Workflows: In-situ Workflows, In-transit Processing





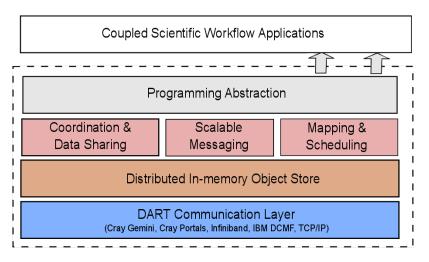
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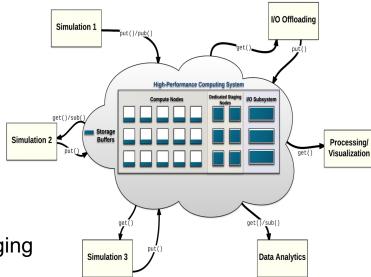




DataSpaces: Data Staging Service for In-Situ Workflows



The DataSpaces Abstraction



- Virtual shared-space programming abstraction
 - Simple API for coordination, interaction and messaging
- Distributed, associative, in-memory object store
 - Online data indexing, flexible querying
- Cross-layer runtime management
 - Hybrid in-situ/in-transit execution
- High-throughput/low-latency asynchronous data transport

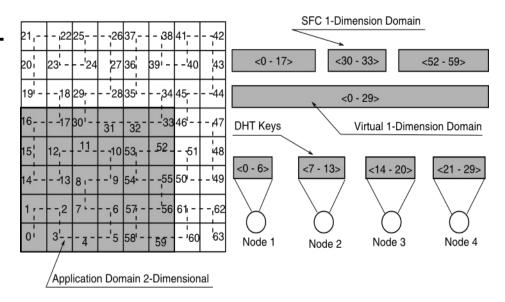




DataSpaces : Indexing + DHT

- Dynamically constructed structured overlay using hybrid staging cores
- Index constructed online using SFC mappings and applications attributes
 - E.g., application domain, data, field values of interest, etc.
- DHT used to maintain meta-data information
 - E.g., geometric descriptors for the shared data, FastBit indices, etc.
- Data objects load-balanced separately across staging cores

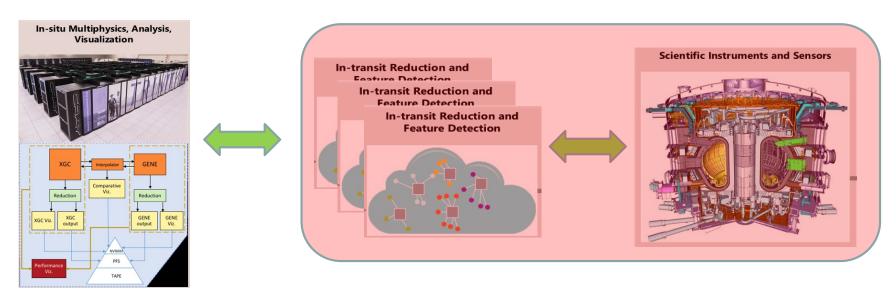




The SFC maps the global domain to a set of intervals

- Intervals are non-contiguous and can lead to metadata load imbalance
- Second mapping compacts the intervals into a contiguous virtual interval
- Split the contiguous interval equally to the DataSpaces nodes

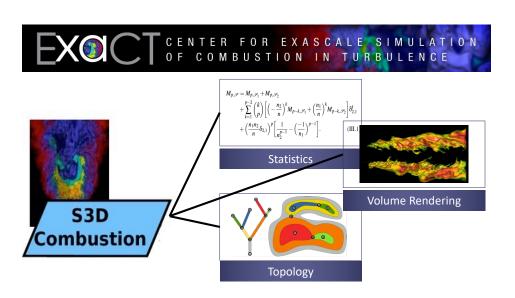
Rethinking Extreme-Scale Workflows: In-situ Workflows, In-transit Processing

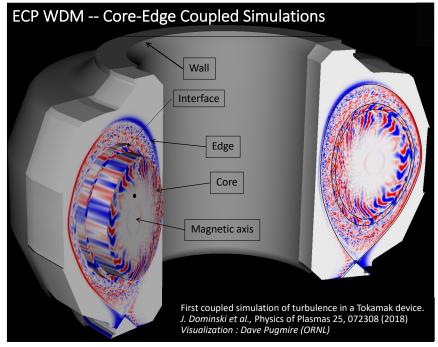


- A. Zamani, D. Balouek-Thomert, J.J. Villalobos, I. Rodero, M. Parashar, An edgeaware autonomic runtime for data streaming and in-transit processing, FGCS, vol. 110, pp. 107-118, 2020.
- M. Aktas, J. Diaz-Montes, I. Rodero, M. Parashar, WA-Dataspaces: Exploring the Data Staging Abstractions for Wide-Area Distributed Scientific Workflows. ICPP 2017, pp. 251-260, IEEE.



Translational Impacts





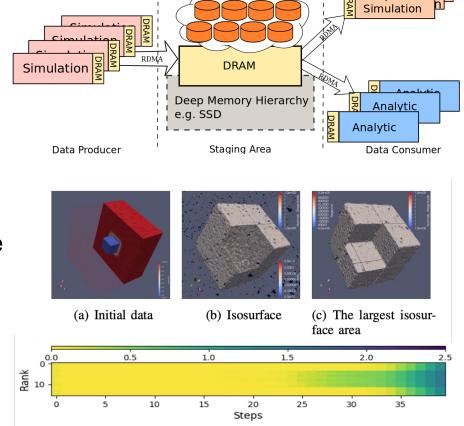




Feedback in Translation: Data Movement, Placement is Important

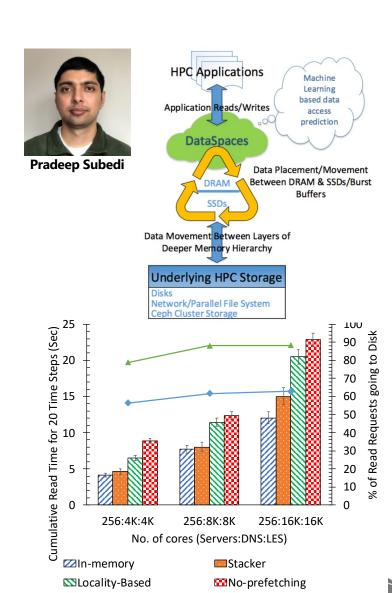
- Storage systems getting deeper and more complex
- Access latencies can severely impact performance
- Data-placement/movement is dependent on the data
- Staging services shared across multiple workflows
- Autonomic, "smart" runtime management needed





Stacker: ML-Guided Data Movement across the Staging Hierarchy [SC18]

- Goal: Reduce the number of read requests to disk using intelligent prefetching; extend in-memory data staging across DRAM, SSDs, burst buffers
- Approach: Data prefetching across the multi-tier staging area using an n-gram model to learn and predict the next data access request
 - Learn inter-application data access patterns, enabling support for multiple concurrent workflows
 - No a priori knowledge of application access pattern or user inputs/hints assumed
- Results: Read performance improved by ~40% as
 compared to no prefetching



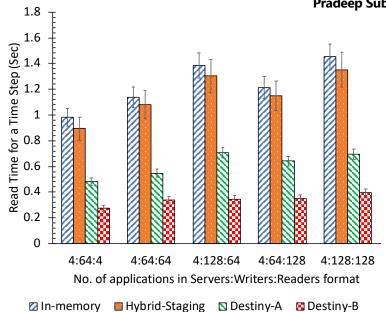
Stacker Disk Access

Destiny: Anticipatory Delivery of In-Staging Data [Cluster 19]



Pradeep Subedi

- Goal: Reduce data access costs for instaging data
- Approach: Anticipate data accesses using an n-gram ML model, and proactively package and move data close to the consumer
 - Amortize expensive data discovery and assembly operations in data staging
- Results: Read response time reduced by up to 75% and 53% for co-located application processes and processes residing in separate nodes respectively



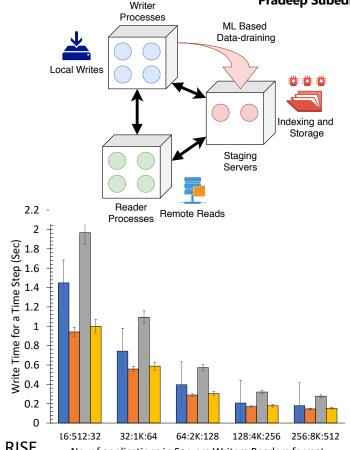
Hybrid-Staging represents just exposing data through shared memory. Destiny-A represents applications residing on separate nodes than staging and Destiny-B represents collocated staging and application processes.

.....

RISE: Reducing I/O Contention for Staging-based In-situ Workflows [Cluster 21]

Pradeep Subedi

- Goal: Reduce data write access costs using "smart" data offloading
- **Approach:** Models the interactions across staging-based coupled applications to predict periods of server idle-time; determine suitable data objects to drained to staging
- Results: RISE captures the data access interval and schedules background drain task, which reduce the write-response time by ~44% as compared to Naive-Drain method and by ~31% as compared to native DataSpaces





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TCS Roadblocks

- In computer science, translation is often confused with commercialization
- Open-source techniques are often confused for translation
- Funding agencies typically don't provide support for translation
 - Resources to sustain and maintain research artifacts (software, data) are essential.
- 4. PhD programs don't allocate time and resources to translation
 - Translation is not considered as an acceptable these topic
- Traditional academic structures, publication venues don't value translation
- 6. There are a lack of exemplars



COVER FEATURE TRANSLATIONAL COMPUTING



There are benefits to formalizing translational computer science (TCS) to complement traditional modes of computer science research, as has been done for translational medicine. TCS has the potential to accelerate





Some thoughts for the path forward...





Funding

- Currently TCS funding is ad hoc => Sustained funding programs and mechanisms focused on fostering and nurturing TCS
- More money alone not solution => Need to build translation into the research plan
- Funding must support experimenting with solutions and prototypes (e.g., Transition to Practice)
- TCS typically involves substantial interaction with end users => Additional funding for travel, user engagement, provisioning of development resources
- Translation process feeds back into the research => May require a loop of research and translation rather than a linear workflow
- Research should be free (encouraged) to report on both research successes, but also translation success or failures







Venues, metrics and reward structures

- Traditional publications are not well suited to TCS => Similar to TM, establish new journals that explicitly target translation
- Metrics, recognitions and rewards structures are lacking, especially in the academic community
 - Mechanisms to track their use, citations and impact are a step in the right direction
 - Metrics that report uptake, and measure how many of these have resulted in successful translation
 - Integrate metrics into promotion processes







Education and Training

- Integration of translational approaches and methodologies into more formal computer science curricula
- New materials and mechanisms for providing translational skills to practitioners, in both computer and other disciplines
 - Institutes/centres have been established that encourage and enable trans-disciplinary research
- Extreme example: A PhD could be entirely devoted to the translation of work performed by another researcher, with no original research on the background IP per se







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Conclusion

- CS research, innovations are transforming science and society
- TCS complements traditional CS research models (foundational, use-inspired, applied) and can accelerate and amplify the impact of computer science research
- There are benefits to formalizing TCS to complement traditional modes of computer science research
 - Several issues: Funding models, reward structures, publication venues, education and training, etc.







Call for Papers: IEEE CiSE Department: Case Studies in Translational Computer Science

Editors:

Manish Parashar, Scientific Computing and Imaging Institute, University of Utah David Abramson, Centre for Research Computing, The University of Queensland

Description: Our new CISE department explores how findings in fundamental research in computer, computational, and data science translate to technologies, solutions, or practice for the benefit of science and engineering, and society. Specifically, each department article will highlight impactful translational research examples in which research has successfully moved from the laboratory to the field and to the community. The goal is improving understanding of underlying approaches, exploring challenges and lessons learned, with the overarching aim to formulate translational research processes that are broadly applicable.





Thank you!

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- David Abramson
- Pradeep Subedi, Philip Davis, Daniel Balouek-Thomert, Zhe Wang, Bo Zhang
- Many students, postdocs and collaborators



