

Towards Autonomic Science Infrastructure: Architecture, Limitations, and Open Issues

Raj Kettimuthu, Zhengchun Liu, Ian Foster, Pete Beckman, Argonne National Laboratory Alex Sim, Kesheng Wu, Lawrence Berkeley National Laboratory Wei-keng Liao, Qiao Kang, Ankit Agrawal, Alok

Choudhary, Northwestern University

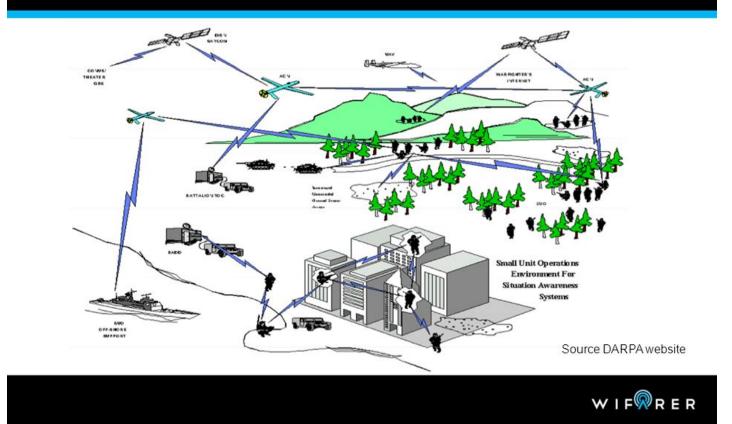


1920 telephony

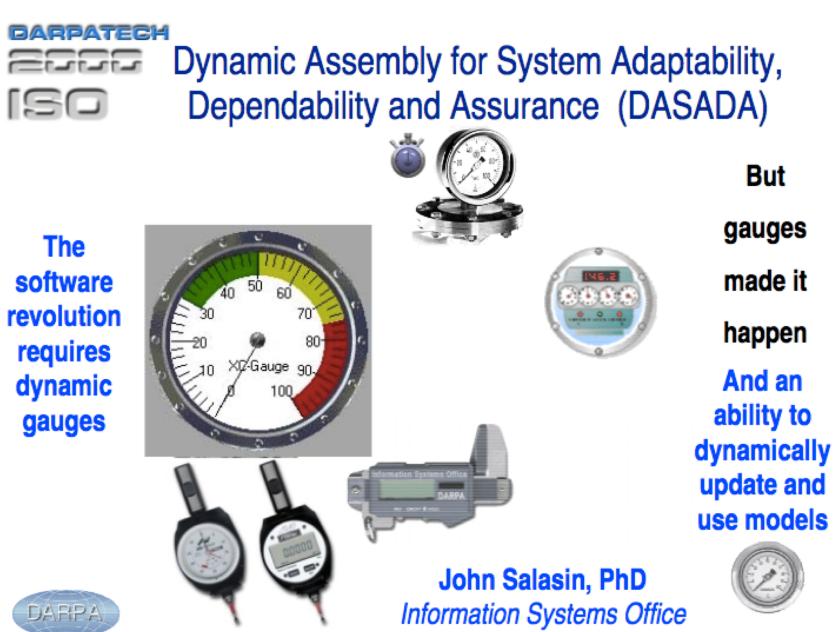


Situational awareness system

SUO-SAS AND Urban fighting



DASADA project

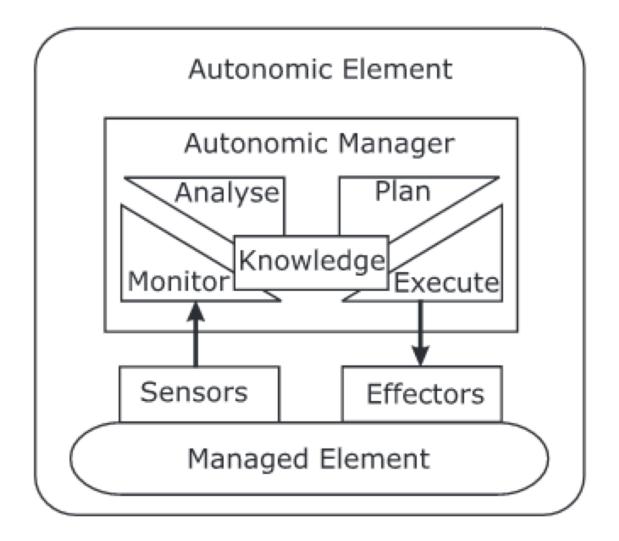


IBM

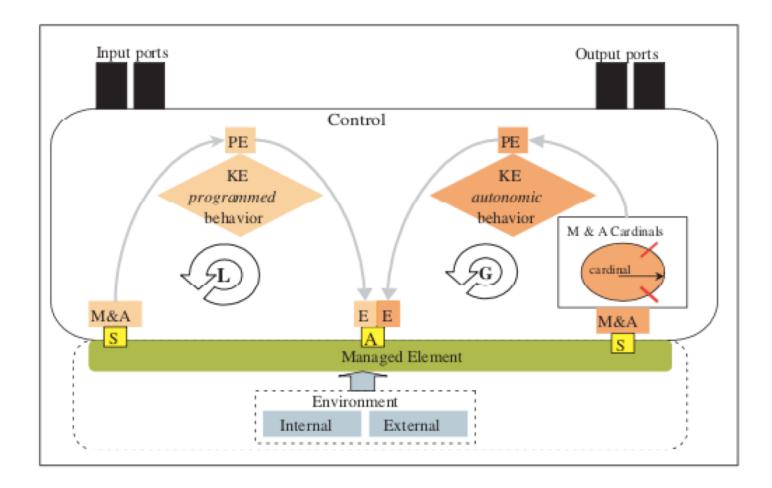
- Complex computing systems should have autonomic properties
- Independently take care of the regular maintenance and optimization tasks
- Reduce workload on the system administrators
- Distilled four properties of a self-managing (i.e. autonomic) system:
 - Self-configuration: adapting to dynamically changing environments
 - Self-optimization: tuning resources and balancing workloads to maximize use of IT resources
 - Self-healing: discovering, diagnosing, and acting to prevent disruptions
 - Self-protecting: anticipating, detecting, identifying, and protecting against attacks



MAPE-K

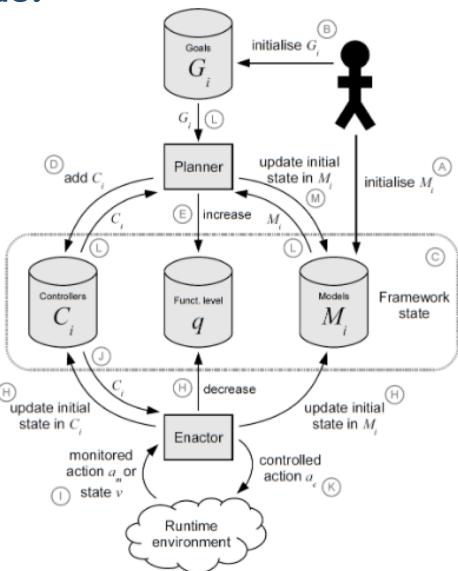


Autonomic element



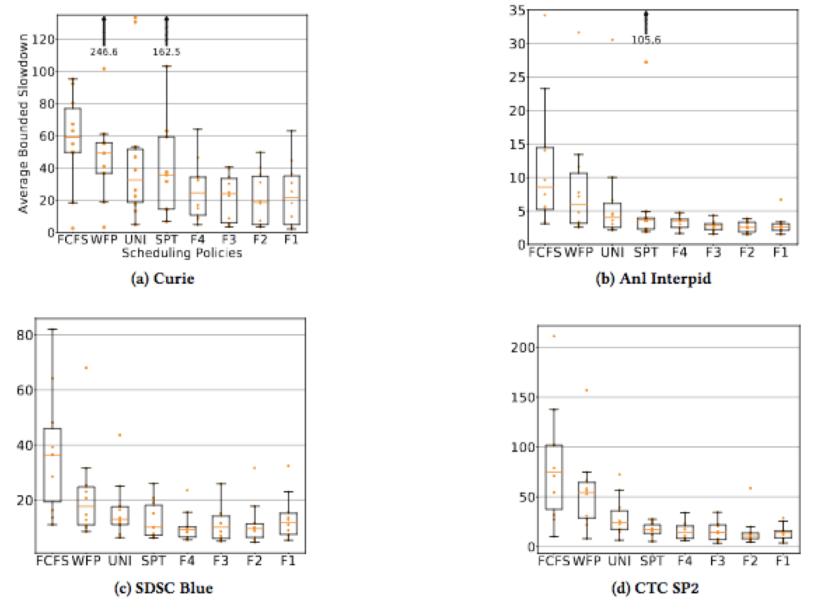
M. Parashar, S. Hariri, "Autonomic Computing: An Overview," Conference Paper in Lecture Notes in Computer Science, January 2004

Multi-tier model



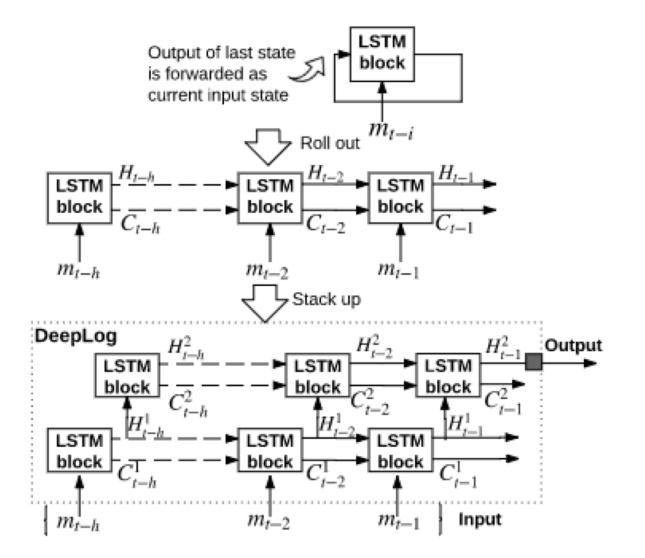
N. D'Ippolito, V. Braberman, J. Kramer, J. Magee, D. Sykes, and S. Uchitel, "Hope for the Best, Prepare for the Worst: Multitier Control for Adaptive Systems," ICSE 2014.

Dynamic job scheduling policies



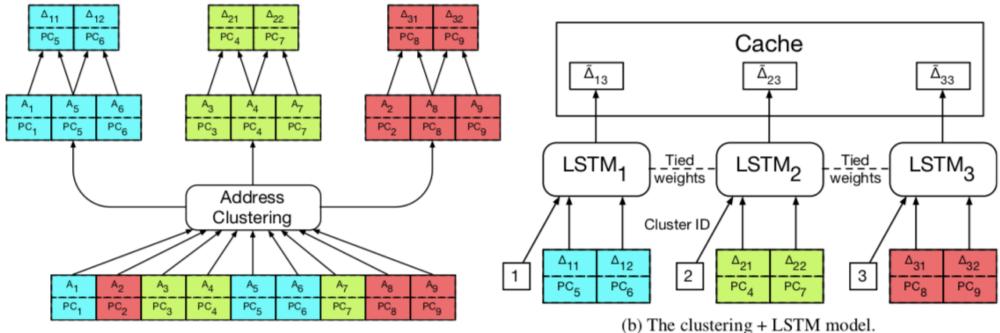
D. Carastan-Santos, R. Y. De Camargo. Obtaining Dynamic Scheduling Policies with Simulation and Machine Learning. SC'17.

Anomaly detection



M. Du, F. Li, G. Zheng, and V. Srikumar, "DeepLog: Anomaly Detection and Diagnosis from System Logs through Deep Learning," In ACM SIGSAC Conference on Computer and Communications Security (CCS '17).

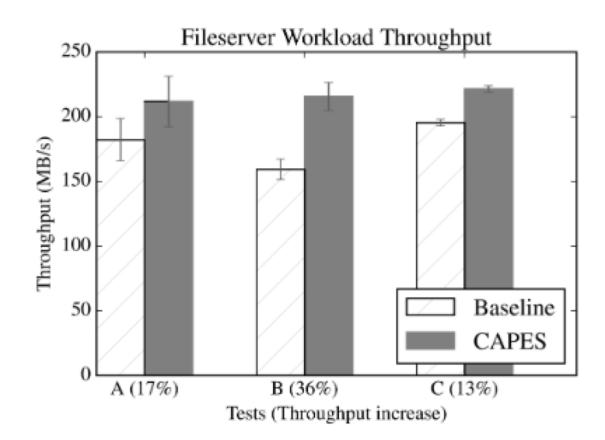
Prefetching



(a) Clustering the address space into separate streams.

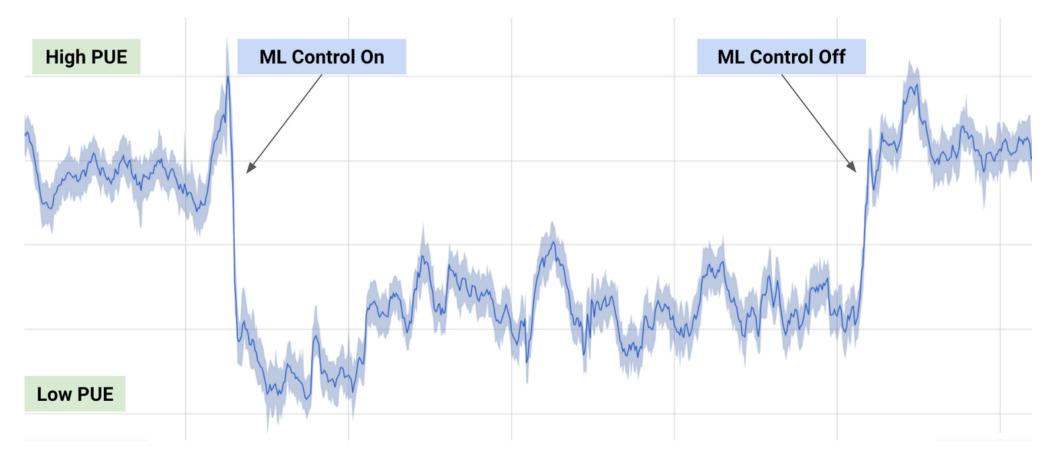
M. Hashemi, K. Swersky, J.A. Smith, G. Ayers, H. Litz, J. Chang, C. Kozyrakis, P. Ranganathan, "Learning Memory Access Patterns", 2018.

Tune performance of storage systems



Y. Li, K. Chang, O. Bel, E. L. Miller, and D. Long, "CAPES: Unsupervised Storage Performance Tuning Using Neural Network-Based Deep Reinforcement Learning," SC17.

DeepMind AI Reduces Google Data Centre Cooling Bill by 40%



https://deepmind.com/blog/deepmind-ai-reduces-google-data-centre-cooling-bill-40/

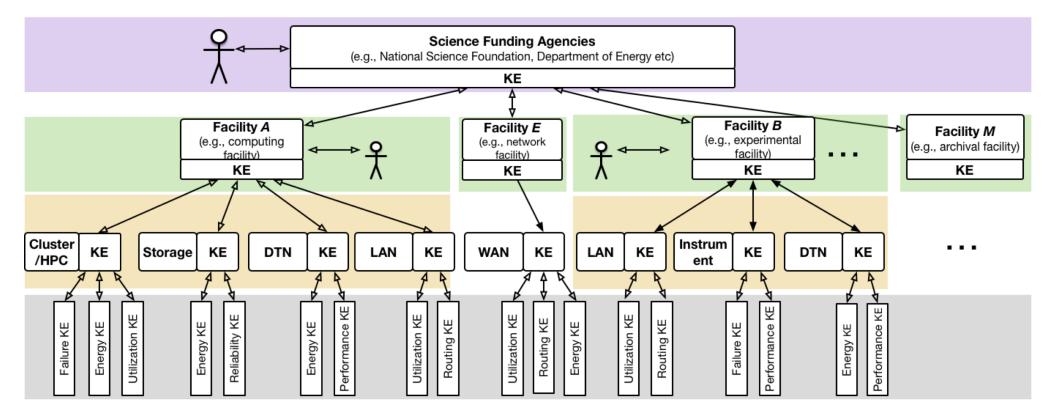
Self driving cars



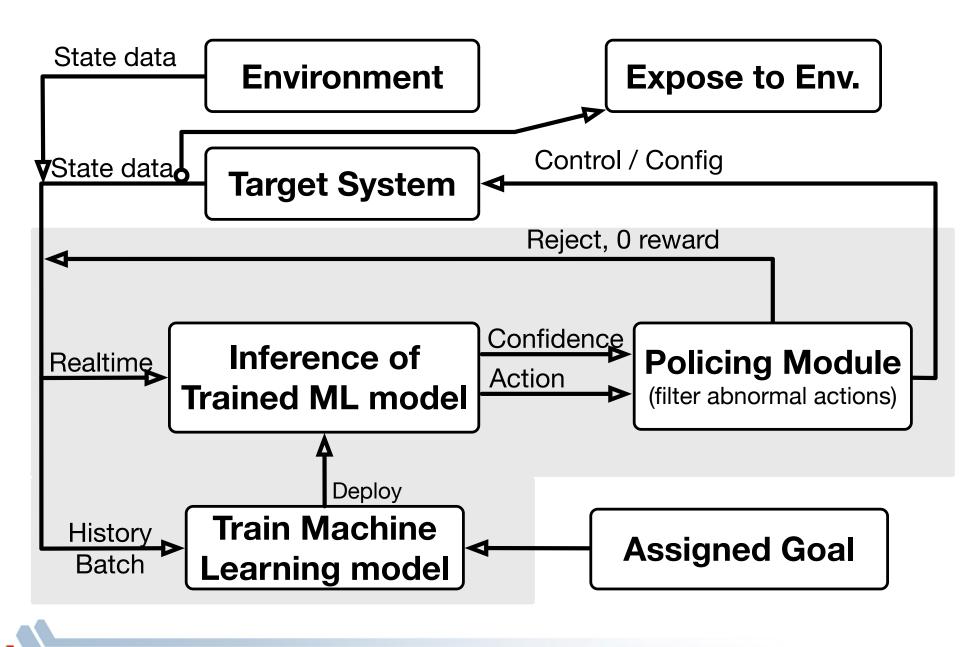
Challenges

- ML methods have promise as a means of creating autonomous modules for computer systems
- Coexistence of several autonomous modules is required in order to handle multiple concerns - requires coordination
- As systems scale toward exascale, many resources will become increasingly constrained
- Some resources have historically been allocated explicitly, others such as N/W BW, I/O BW, and power are not
- As systems continue to evolve, we expect many such resources will need to be explicitly managed
- Autonomous management of resources and coordination to achieve higher-level goals even more important

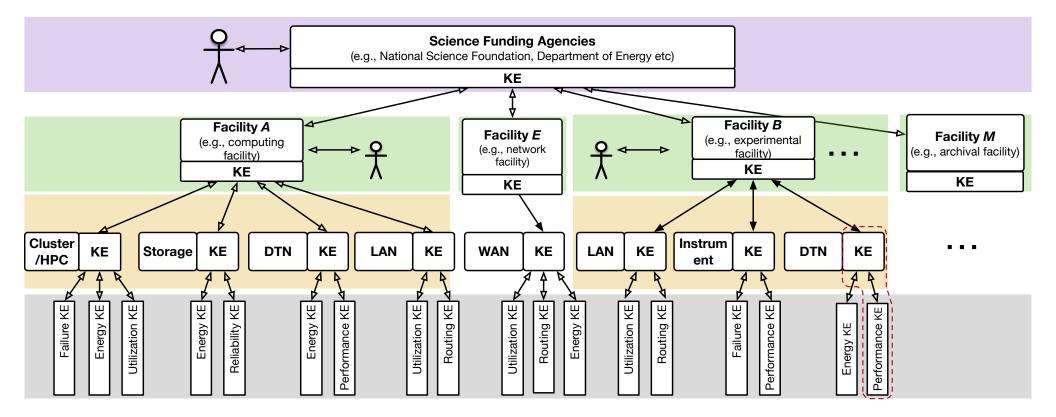
Autonomous science infrastructure architecture



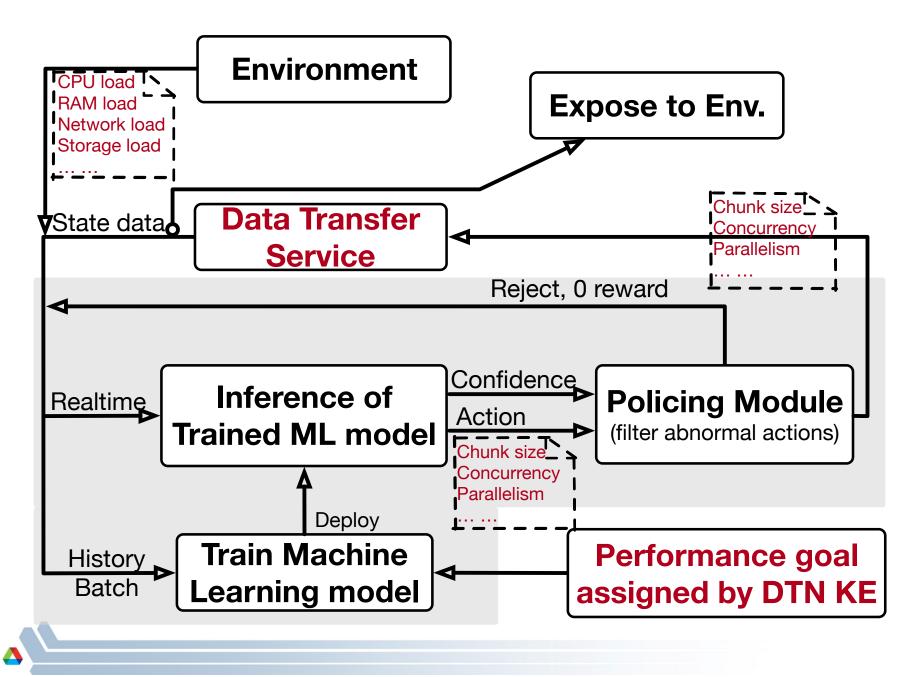
Architecture of an individual autonomous system



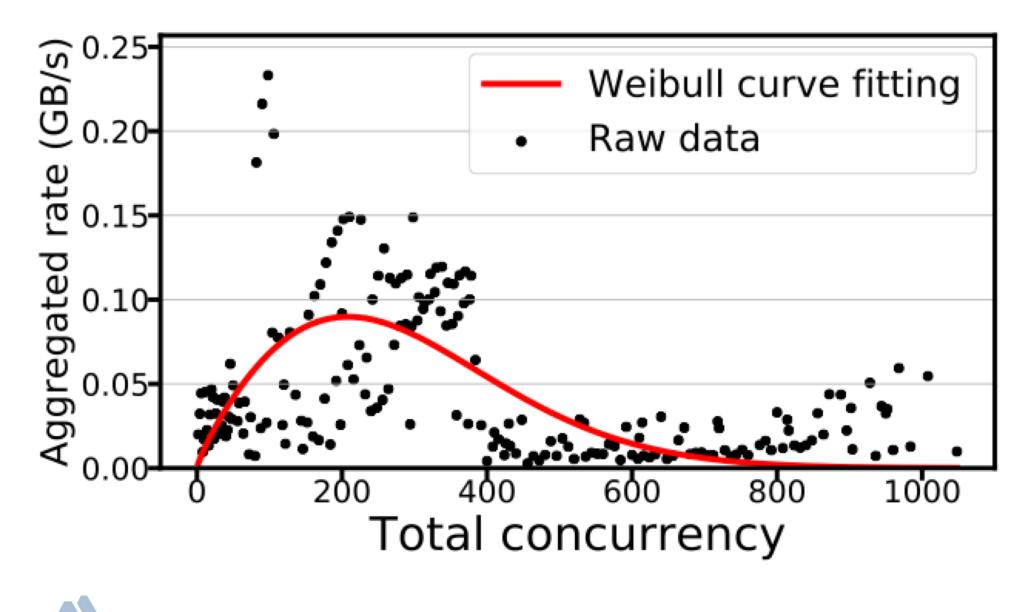
Autonomous science infrastructure architecture



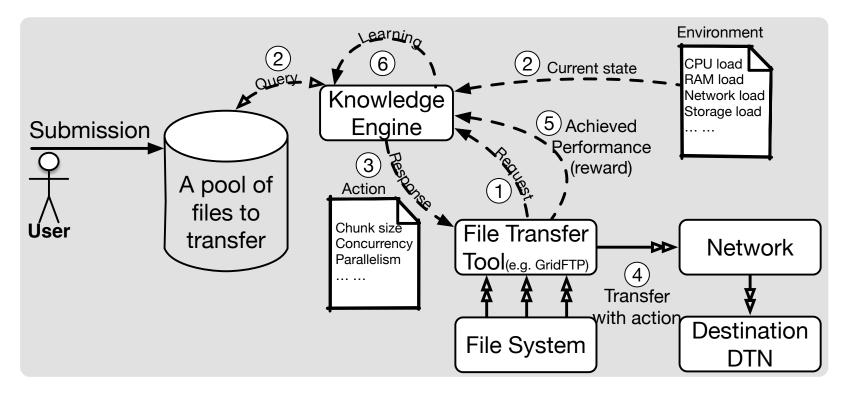
Autonomous performance module for DTN



Concurrent transfers

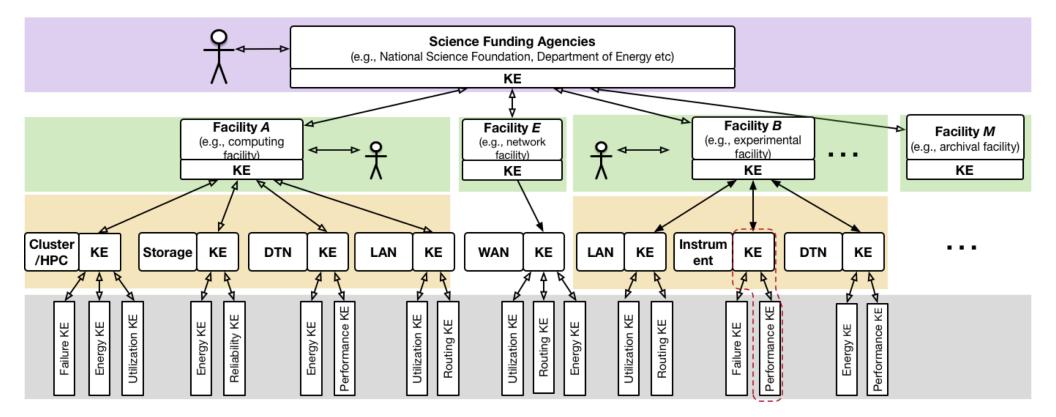


Smart data transfer node

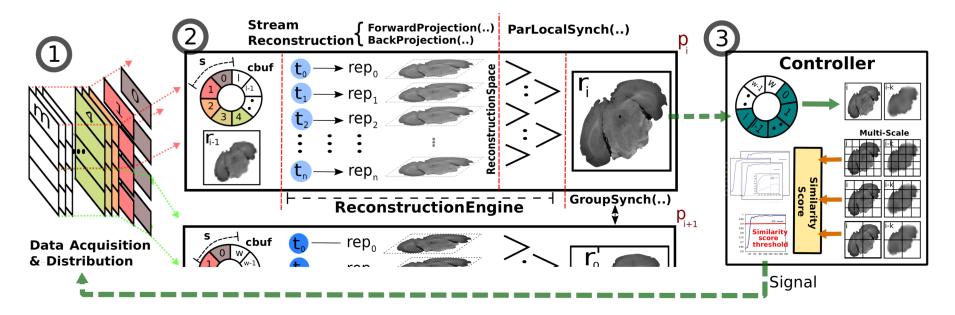


(1) A file transfer tool requests a file to transfer from the KE. The KE (2) checks the current DTN state and (3) responds to the transfer tool with a chunk of file and corresponding optimal transfer parameters (the steering action). (4) The transfer tool transfers the associated chunk with the parameters and monitors the aggregate DTN throughout during this transfer. (5) Once completed, DTN's average aggregate throughput is reported to the KE as a reward for its actions. (6) Based on the reward (encourage or discourage), the KE updates its internal model parameters to improve its decision policy.

Autonomous science infrastructure architecture

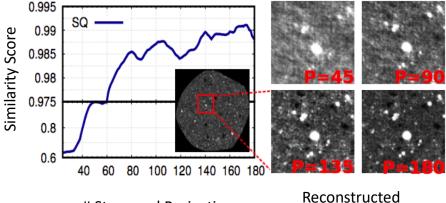


Experiment monitoring and steering



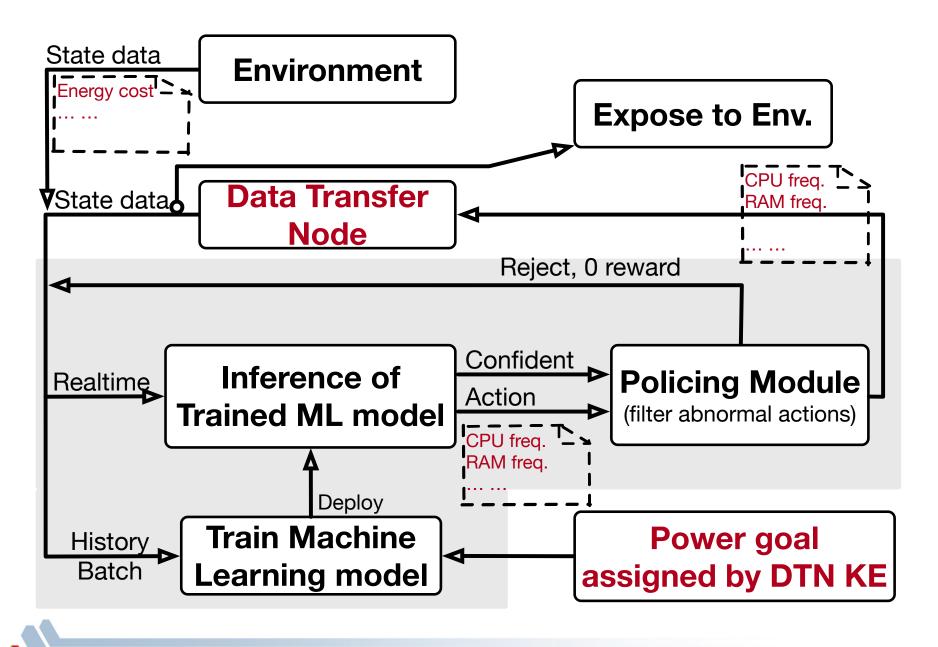
Autonomous stream processing system that allows data streamed from beamline computers to be processed in real time on a remote supercomputer with a control feedback loop used to make decisions during experimentation

 Reduce data acquisition time by 22–44% for the datasets considered in our experiments

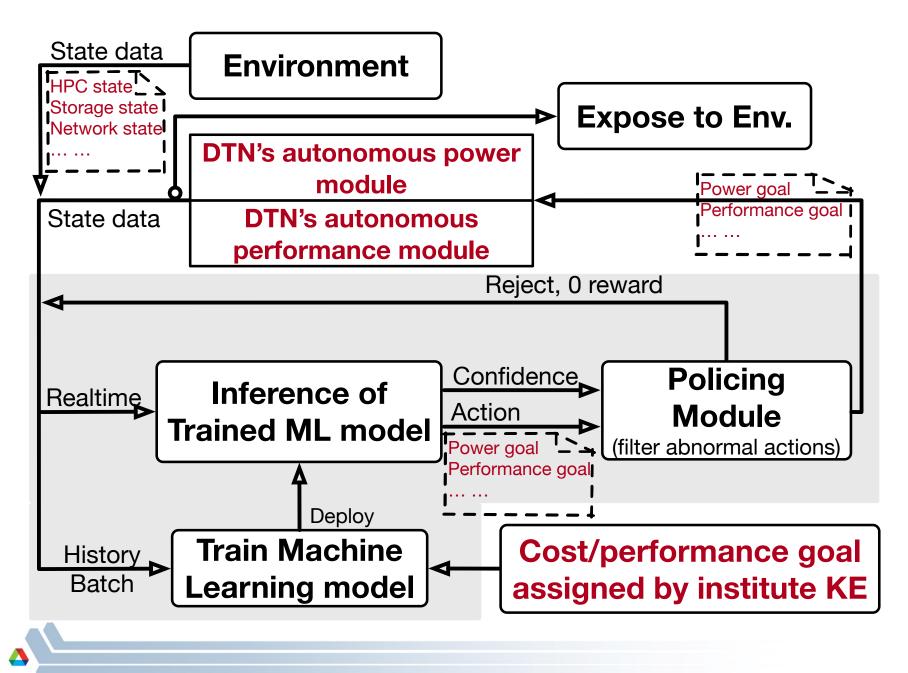


Streamed Projections Image Sequence Image Quality with respect to Streamed Projections

Autonomous power module for DTN



Autonomous DTN



Questions

