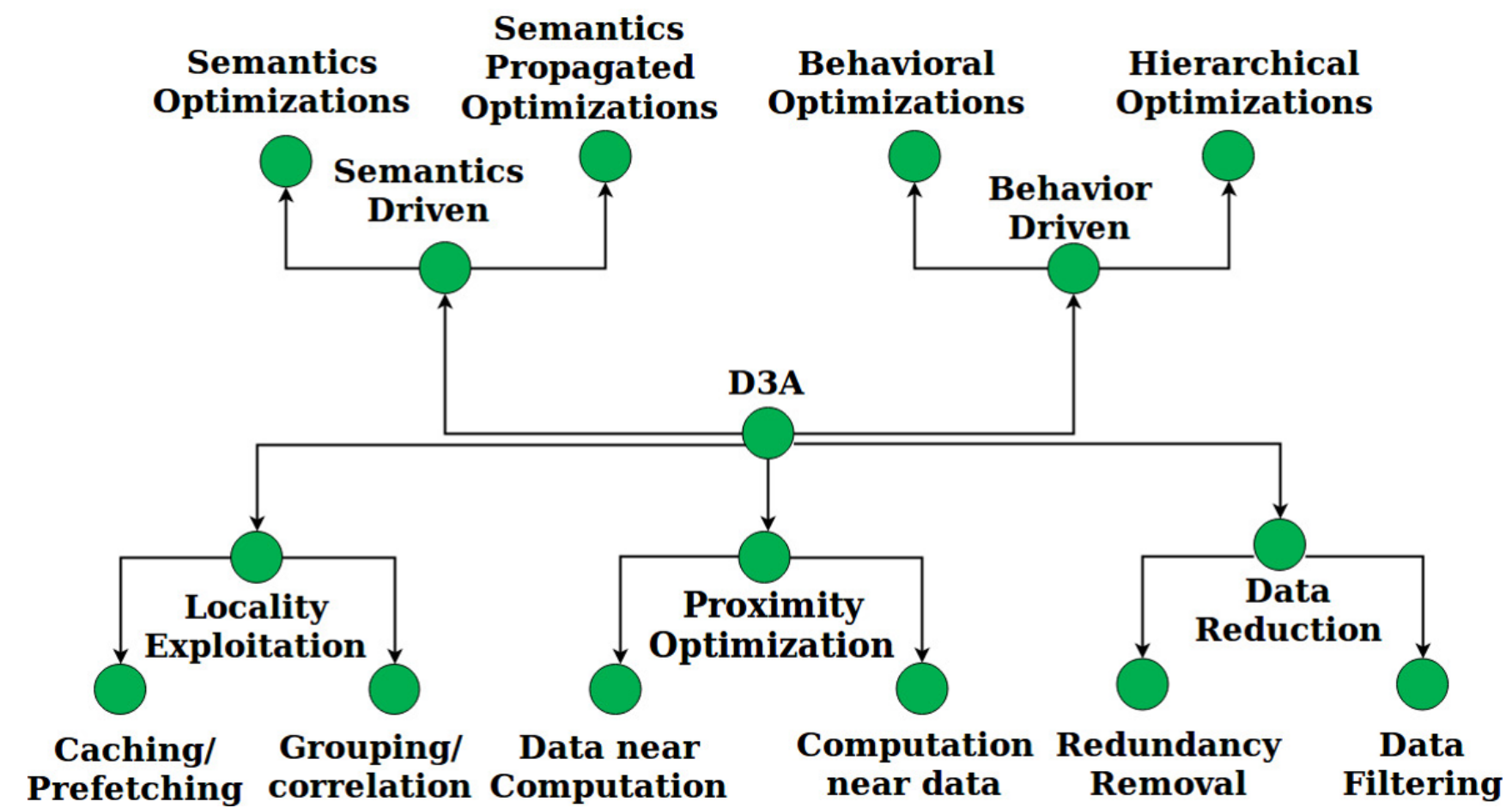


Motivation

- Data size already enormous and growing:
 - Over 90% of world data generated in last 2 years
- Large scale systems highly distributed
 - Resource sharing brings opportunities and challenges
- Strain and opportunity for data centers and cloud storage providers
 - Efficient pattern discovery and description
 - Observed predictability of complex patterns
- Intelligent management of storage tiers is crucial

Data Driven high performance Data Access (D3A) overview



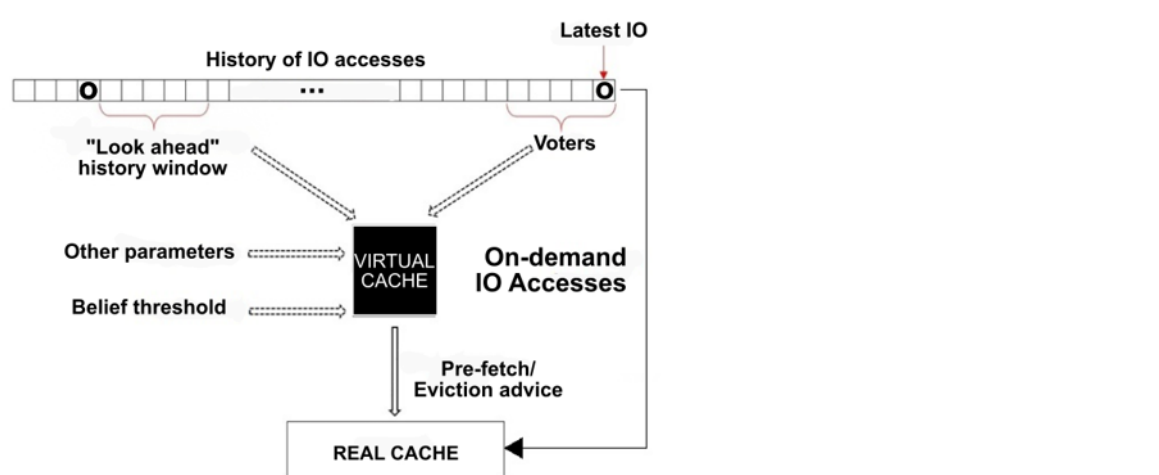
Locality Exploitation Foundations

- Belief – estimate of a probability of a hit
- Conditional probability x being in window of calculated:

$$P(\{x\} | E = o) = \frac{c_{o-x}}{c_{x_p}} * \frac{f_x}{f_o}$$

- $\{x\}$: x is in the “look-ahead” window
- $\{x\} | E = o$: $\{x\}$ conditioned by current object o
- c_{o-x} : count of x in “look-ahead” windows of o
- c_{x_p} : count of x in all “look-ahead” windows
- f_o, f_x : frequency of o , and x respectively

- Modular approach allows for
 - Decoupling decision making from cache management
 - Reducing the computational overhead of belief computation

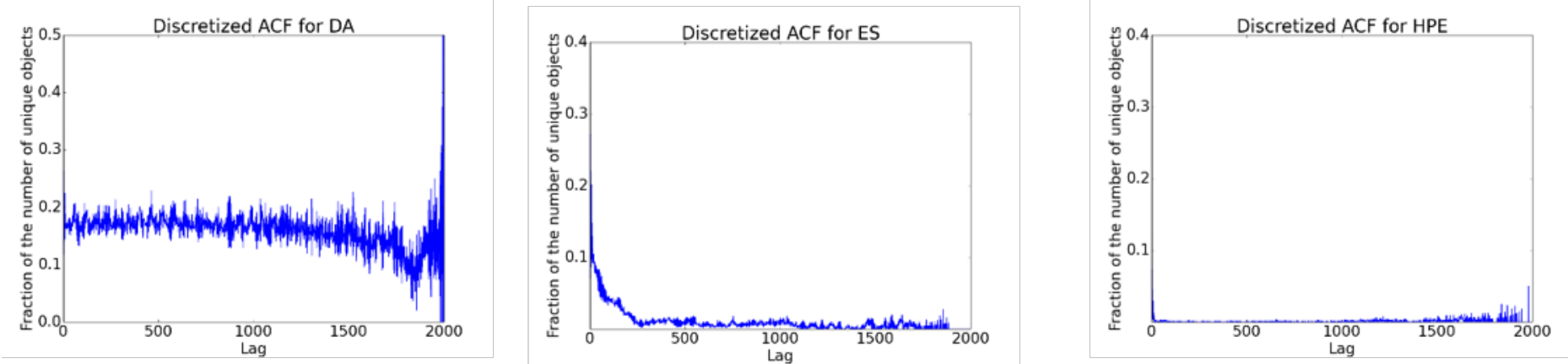


- Framework exploits
 - Various levels of successor predictability
 - Indirectly exploits access history usage patterns
 - Local/global view combination expected to provide best decisions

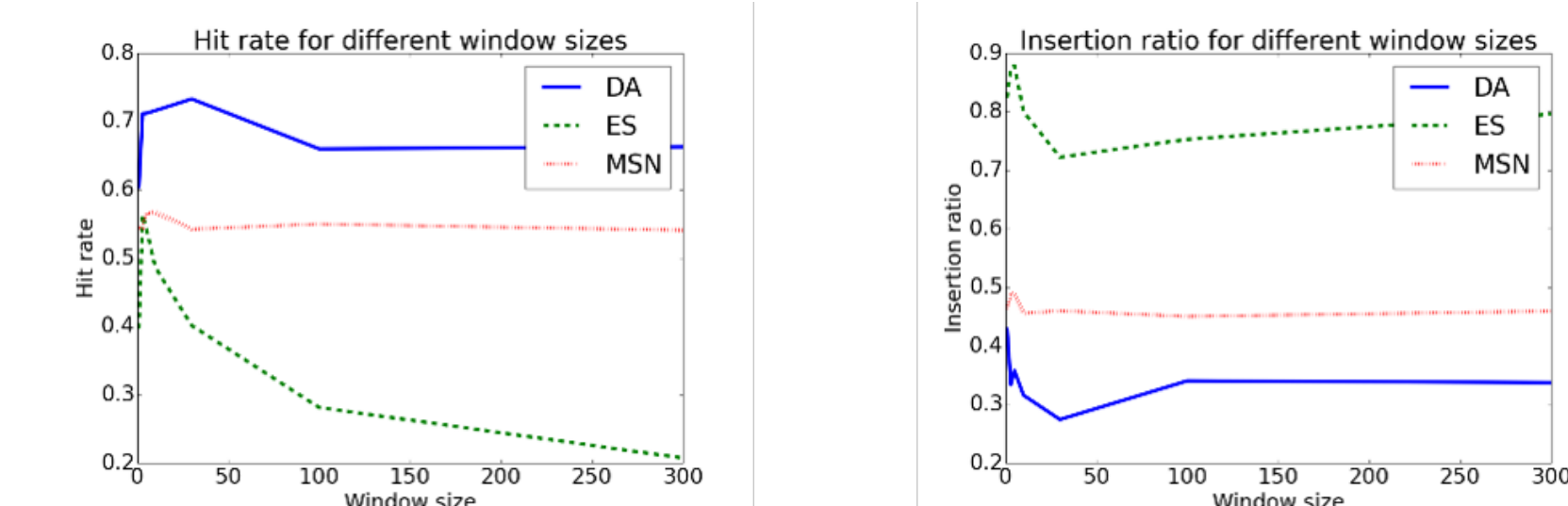
- Obtaining good results largely depends on parameters
 - For the perturbations of the same workload able to adapt
 - For the drastic change in workload/different workload - retrain

Locality Exploitation Evaluation

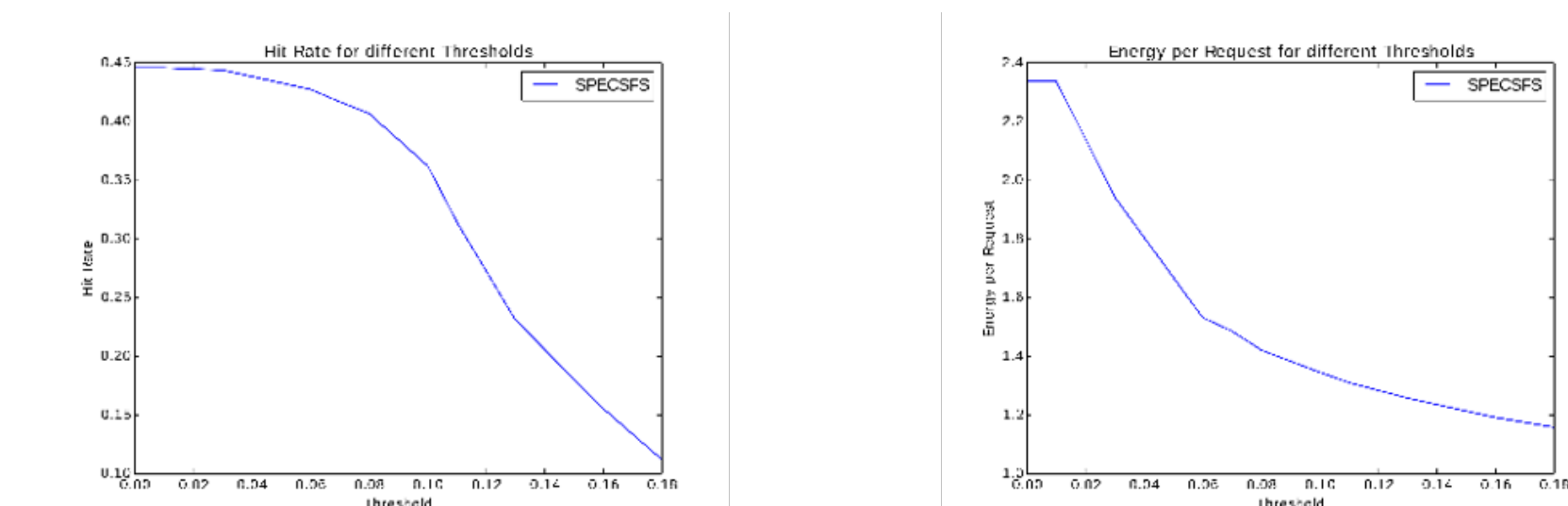
- Microsoft block traces (available at SNIA website)
 - Primarily Disk IO, FileIO traces with following unique characteristics



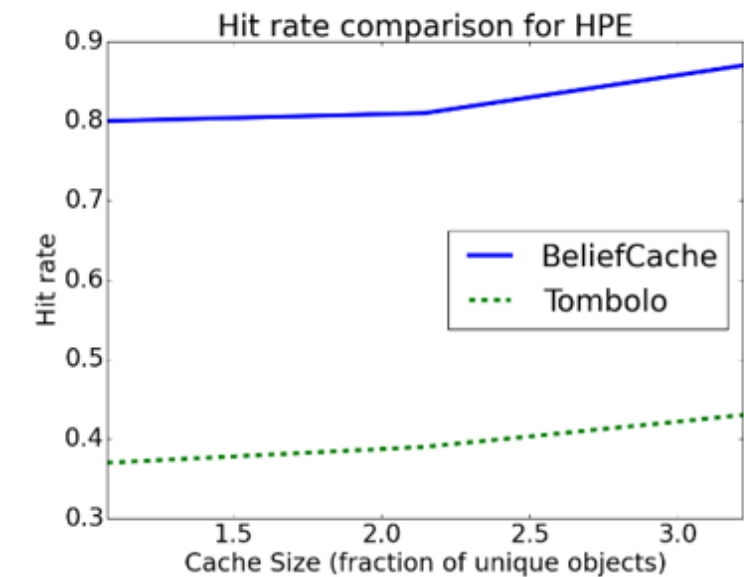
- SPEC SFS suite
 - Video Data Acquisition workload (HPE above)
 - Virtual Desktop Infrastructure workload
- Noticeable trends
 - For workloads exhibiting short and medium term correlations



- For workloads exhibiting long term correlations



- Complex non-sequential patterns captured better than state-of-the-art framework for optimizing cloud storage gateways



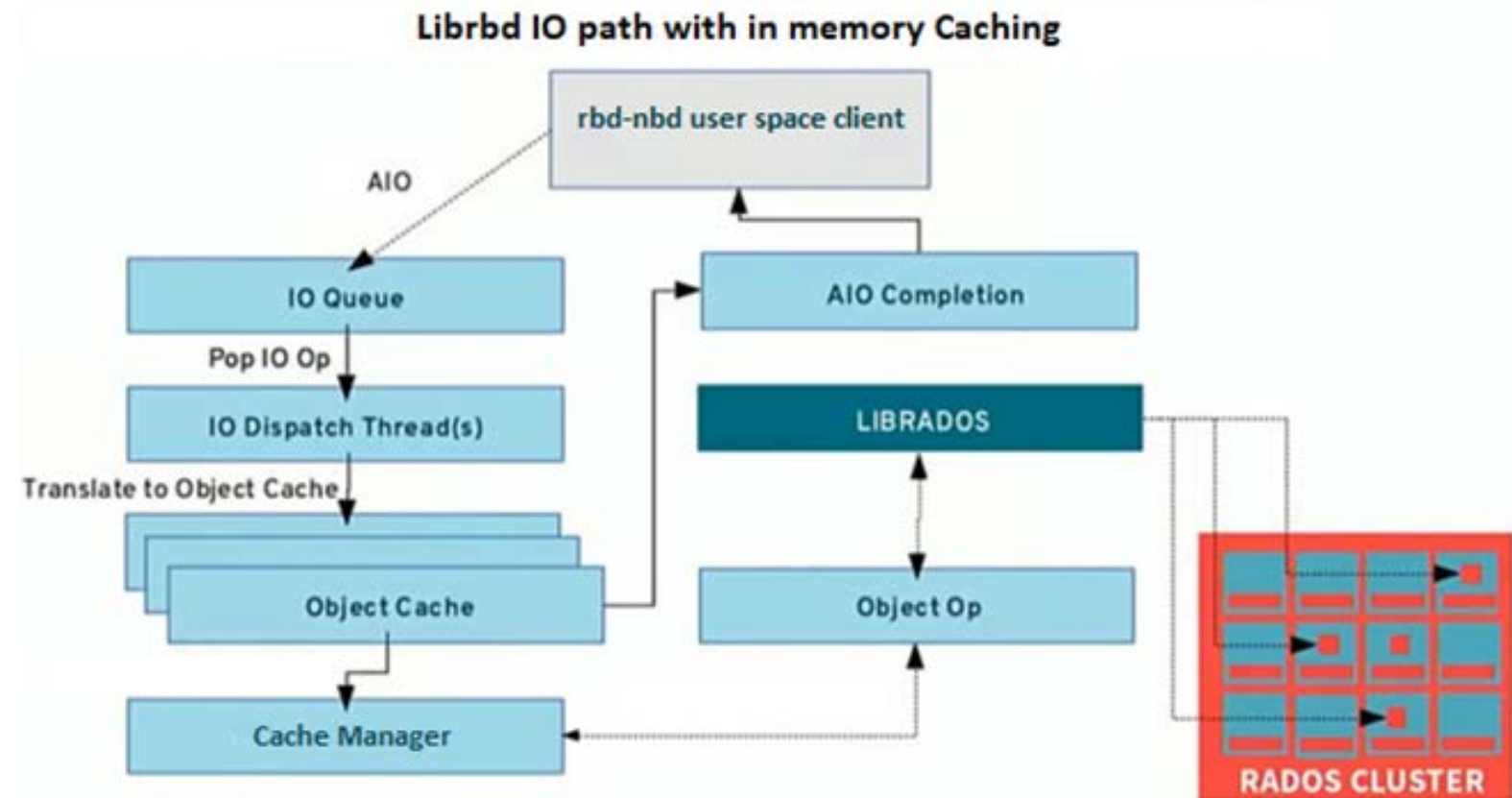
- Further improvements are possible by
 - Understanding the effects of phase change on caching characteristics
 - Using the phase change detection mechanism to automate retraining

- Optimized stripe cache splitting between DRAM and NVRAM
- Balance trade-off between Performance, Resilience, Endurance
- Built efficient, flexible mechanism that adapts to
 - Amount and speed of DRAM and NVRAM
 - Workload Characteristics (VDI trace on picture below)
- IC – instance of locality exploitation evaluation
- RH – instance of currently applied policies
- Perfect – what we could get if we knew all the information



Locality Exploitation Implementation

- Previous prefetching
 - Read ahead of extra bytes from same object upon sequential access
- Challenges
 - Keeping track of object id is not enough
 - Requires asynchronous requests
 - No fixed unit of request
 - Entire object may be too big to cache
 - Complex striping



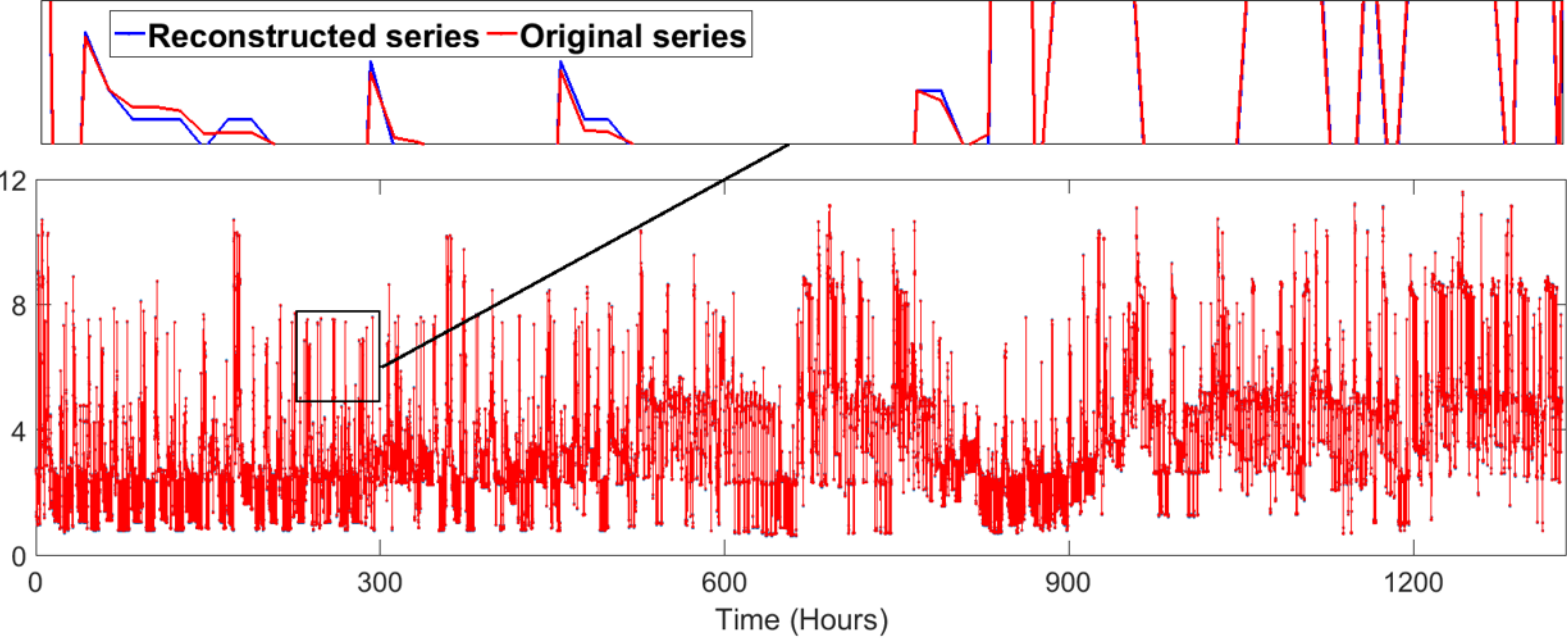
Data Reduction Foundations

- The increasing proliferation of the IoT devices and systems
- Large amounts of highly heterogeneous data to be collected
 - Multiple sensors “readings” at successive time points
 - A continuous stream of data is being generated
 - With n sensors each sample will be an n -element vector
 - Use cases when it is acceptable to scale the received values
 - Discretize the value of the i -th sensor into b_i bins, for a suitable b
- Provenance information is required
 - Describe the data in a sufficiently intuitive way - easily usable
- Lossy data compression mechanisms necessary
 - Provide a trade-off between
 - Compression ratio and
 - Data usefulness for offline statistical analysis
 - Preserve the information about the sequence of events
 - Crucial in IoT context

- Ensure that the sequence of events can be derived (or inferred) from the compressed data
- Keep some events intact and approximate other events to improve the compressibility
- Two stages of the method
 - Use the SQS1 to generate a summarization of the data S
 - Approximately match the remaining events with the patterns in S
 - A matching threshold τ is the Euclidean distance between the subsequence and the existing pattern from S
- We consider the following two ways to apply SQS:
 - (Correlated AVSC-C) Directly compress each vector
 - Retains the correlation across streams perfectly,
 - It may show poor tradeoff of compressibility and accuracy
 - (Uncorrelated AVSC-NC) Compress i -th stream separately
 - Different approximations across vectors might lead to approximate information about corresponding streams
 - Expected to yield better compression at the cost of confounding the correlation across the attributes

AVSC Discussion

- Lossy compression shows gains over the plain lossless compression
 - For a specified amount of accuracy for purposes of identifying the state of the system
- Trade-off in between the compressibility and fidelity
 - With respect to distance threshold τ
 - Fidelity increases with small thresholds τ at the cost of poor compressibility
- Observed behavior is domain specific
 - There is a strong correlation between weather and energy usage
 - Discretization smoothed the differences between the vectors



Conclusions and Future Work

- Opportunities and challenges in developing data driven high performance data access methods
 - Leverage data properties and relationships between data items
 - Minimize data movement, latency, energy consumption
- Rapidly emerging data intensive applications should focus on
 - Proximity optimizations leading to reduction of distance between data and computations
 - Data reduction by removing redundancy in data
 - Using sparse representations of data
 - Impact of data access mechanisms on
 - Resilience
 - Energy consumption
 - Storage usage
 - Enablement of new classes of data driven applications

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