UMassAmherst

PRESTO: Feedback-driven Data Management in Sensor Networks

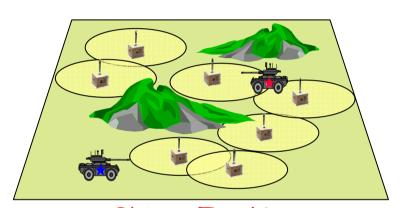
(*PREdictive STOrage)

Ming Li, Deepak Ganesan, and Prashant Shenoy University of Massachusetts, Amherst



Emerging large-scale sensor networks

- Wired/wireless network of spatially distributed sensing devices
- Closely and densely monitor physical phenomena



Object Tracking



Ecosystem Monitoring



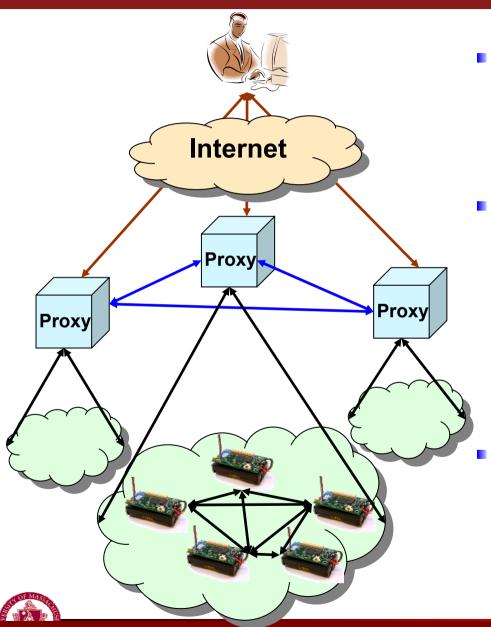
Habitat Monitoring

Resource Constraints

- Energy -> 1 AA battery for 1 year
- Computation < Storage << Communication</p>



Hierarchical Sensor Network Architecture



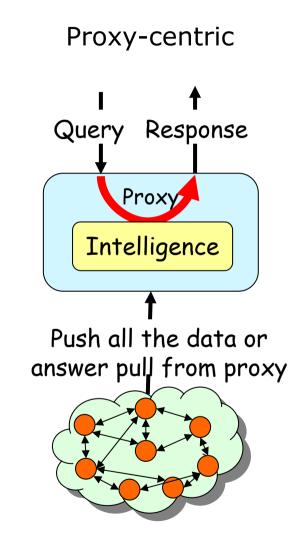
- User
 - Issues queries to the proxy.

- Proxy
 - Gateway between user and sensor
 - PDA class microcomputer
 - Tethered or solar-powered

- Sensor
 - Sensing physical world
 - Resource constrained

Approaches to Proxy-Sensor Interaction

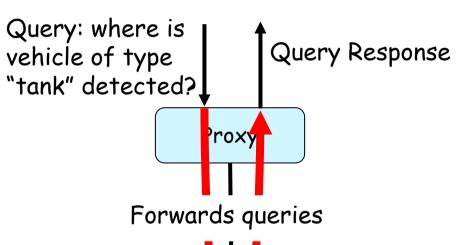
Sensor-centric Query Response Proxy Forwards queries into sensor network Intelligence





Sensor-Centric Architecture

Directed Diffusion [Heidemann01]



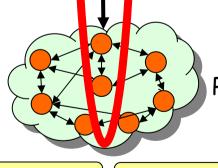
 Sensors perform data fusion, query processing and filtering within the network.

Pros:

- Greater query accuracy
- Better energy-efficiency.

Cons:

- Greater sensor complexity.
- Greater query latency.



Pub-sub routing

Query Process

Data Fusion

Pub-Sub

Storage



Proxy-Centric Architecture

 Proxy processes queries and uses complex spatial/temporal model to predict

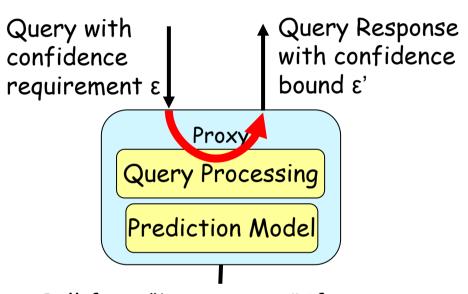
Pros:

- Intelligence placed where resources are available.
- Lower query latency

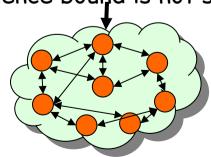
Cons:

- Hard to detect anomalies.
- Sacrifice energy-efficiency or query accuracy.

BBQ [Deshpande04]



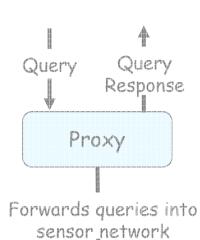
Pull from "best sensor" if query confidence bound is not satisfied.

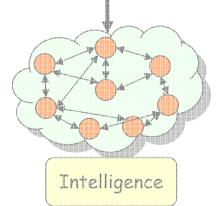


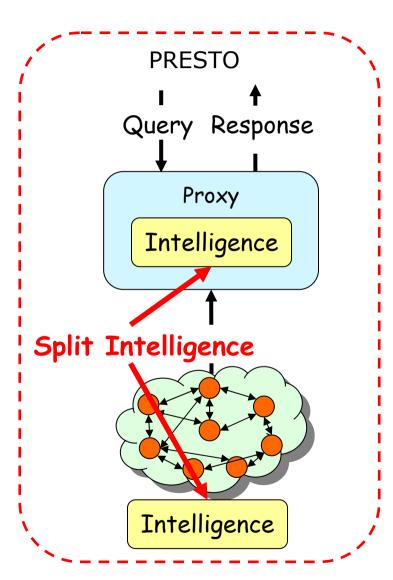


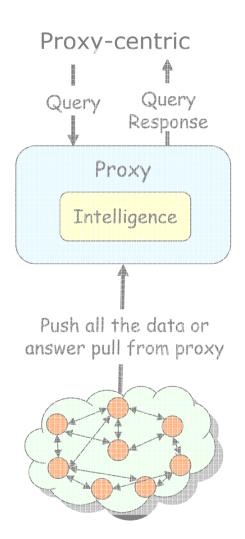
PRESTO

Sensor-centric











Key Ideas in PRESTO

Split intelligence between the proxy which has greater resources, and the sensor which has ground truth.

Exploit predictable structure in sensor data

 Use continual feedback from the proxy to sensors to adapt to data and query dynamics

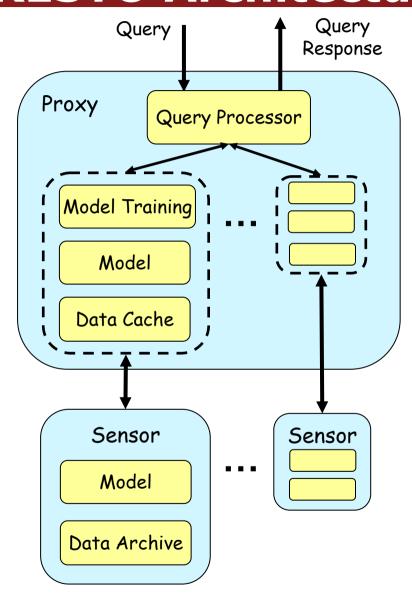


Outline

- Motivation & Key Ideas
- PRESTO Architecture
- ARIMA Model
- Evaluation
- Summary & Future Work



PRESTO Architecture



Query

Which sensor to query

Which time point to query

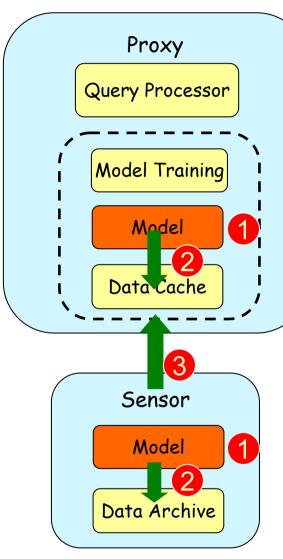
Confidence bound of the result

Key Components

- Model-driven Push
- Model-driven Pull
- Cache Update
- Adaptive Feedback



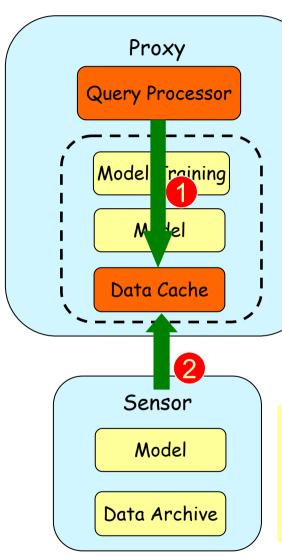
Model-driven Push



- Capture the unpredictable data to guarantee the accuracy of the prediction.
- Procedure
 - In each sampling period, sensor and proxy both predict current data
 - 2. Proxy and sensor cache the result. Proxy also caches the confidence of the prediction
 - 3. Sensor pushes data if difference between the prediction and the real data is higher than a threshold
- Only data anomalies are transmitted
- Bound the error of cache



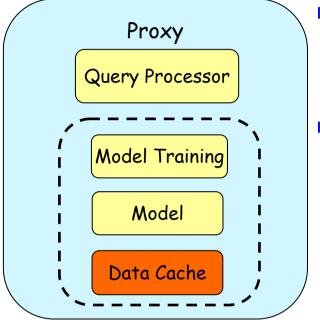
Model-driven Pull



- Answer the queries which can not be satisfied by the cached data
- Procedure
 - Proxy answers query using cached data if it is accurate enough
 - Proxy pulls data from sensor if the cached data does not satisfy the query
- Ensure the accuracy of the query response
- Most of the queries can be answered using cached data.



Cache Updating

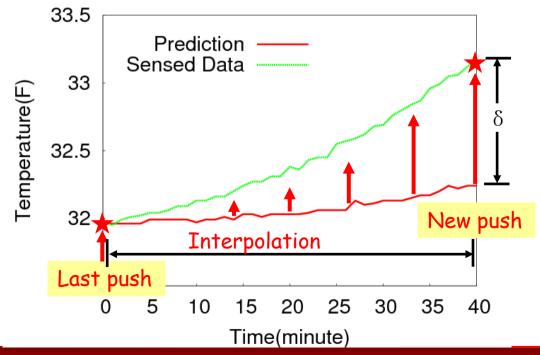


Sensor

Model

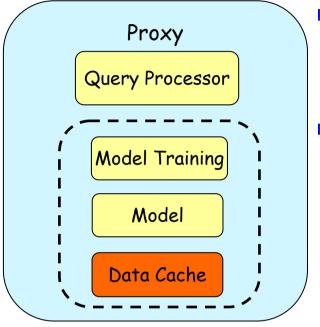
Data Archive

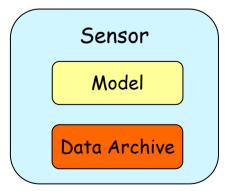
- Use any information from sensor to refine the accuracy of data cache
- Cache is refined after each push/pull
 - Interpolation
 - Re-prediction



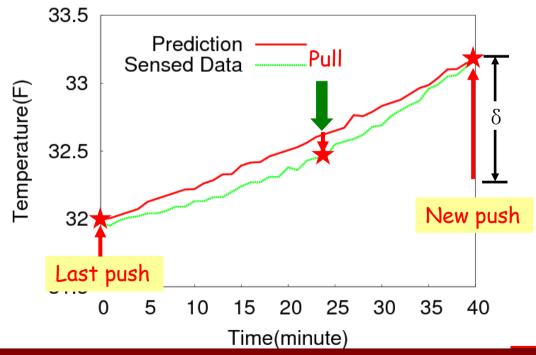


Cache Updating



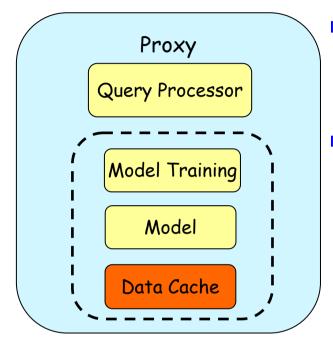


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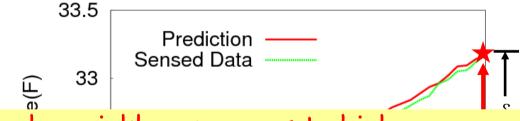




Cache Updating



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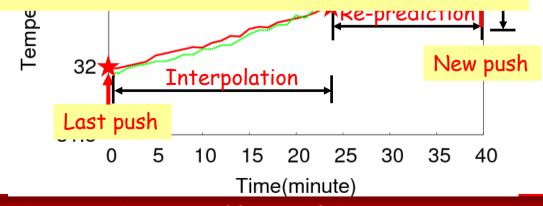


Sensor

Model

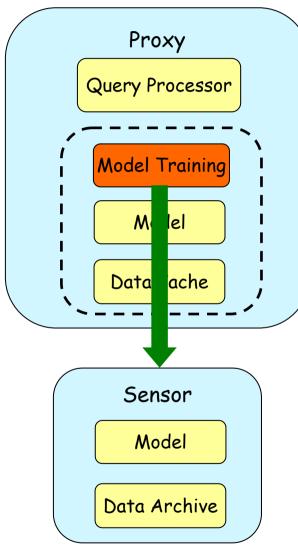
Data Archive

Cache quickly converges to high accuracy



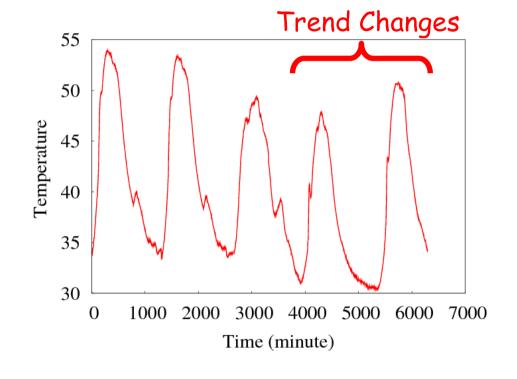


Adaptive Feedback



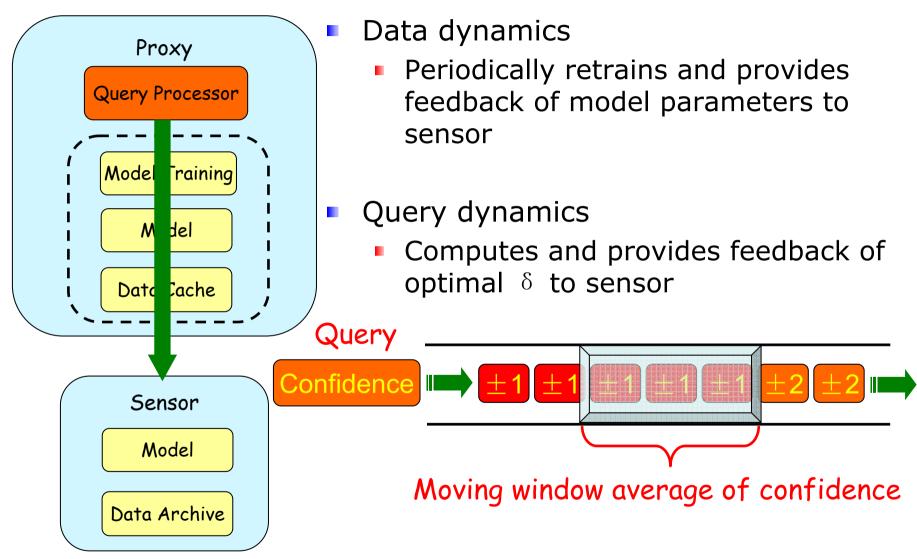
Data dynamics

 Periodically retrains and provides feedback of model parameters to sensor





Adaptive Feedback



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Split Computation Load

 Allocate computation between proxy and sensor according to their resources.

Auto-Regressive Integrated Moving Average Models (ARIMA)

Training/Re-training at Proxy

Min Mean Square Error (MMSE) Estimator requires thousands of floating point operation for training

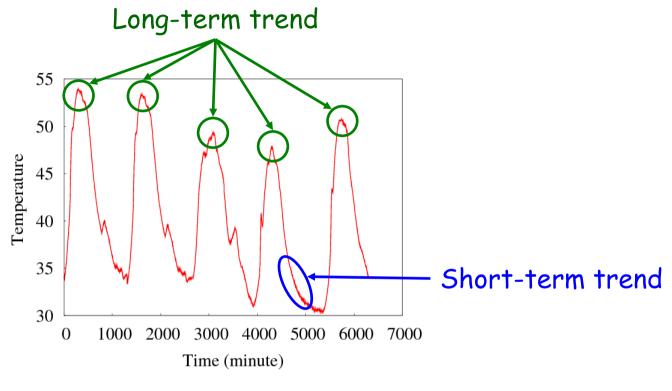
Simple Prediction at Sensor

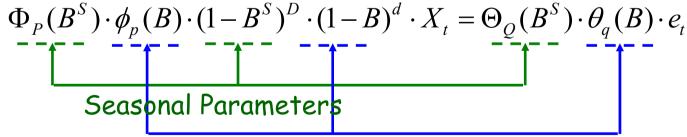
Three multiplications and five additions

$$X_{t} = X_{t} + X_{t-S} - X_{t-S-1} + \theta e_{t-1} - \Theta e_{t-S} + \theta \Theta e_{t-S-1}$$



Capture Long-term and Short-term Trends





One Step Parameters



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Evaluations

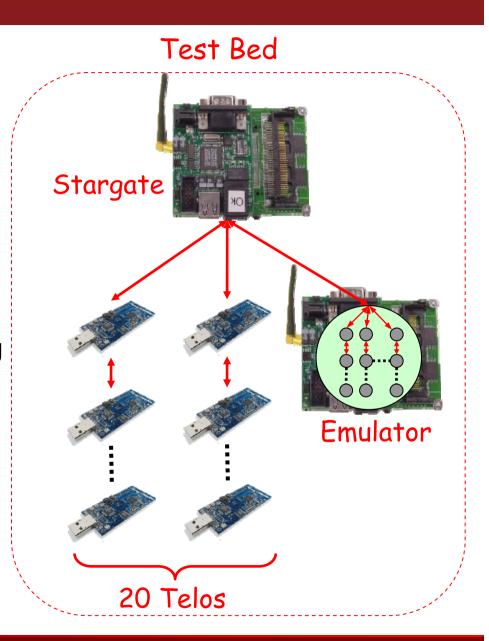
 Uses both numerical simulations and real deployments

Datasets

- One month of temperature readings from from James Reserve at CENS, UCLA.
- Live temperature monitoring at Umass Amherst.

Metrics

- Number of messages
- Query latency
- Mean-square error





Sensor Micro-Benchmark



Breakdown of Energy Costs

Component	Operation	Energy (nJ)
Model	Predict 1 Sample	27
Data Archive	Read + Write + Erase 1 sample	21
Push/Pull	Transmit 1 sample + Receive 1 ACK	3300

- PRESTO exploits modeling and archival to save the cost of pushes and pulls.
- Prediction operation at the sensor incurs low energy cost

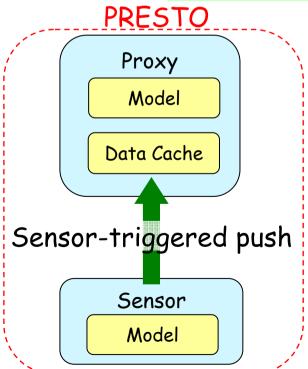


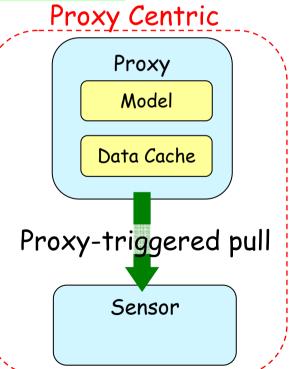
PRESTO vs. Proxy Centric

Compare energy efficiency and cache accuracy of PRESTO against proxy-centric approach

Setup

- Matlab simulations
- Real temperature trace
- Without queries

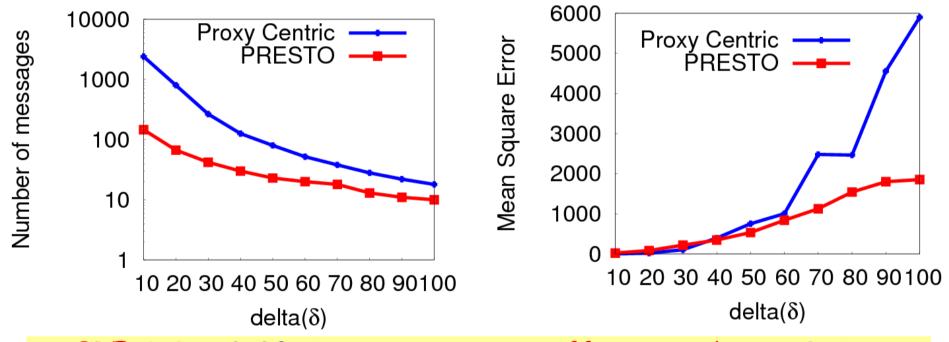






PRESTO vs. Proxy Centric

Compare energy efficiency and cache accuracy of PRESTO against proxy-centric approach



PRESTO is 2-20 times more energy efficient and up to 3 times more accurate



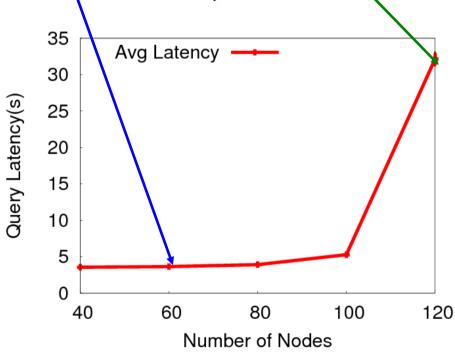
Scalability

How many sensors can a single proxy support at high query rate?
Congestion point of PRESTO

Congestion point of sensor-centric system

Setup

- Runs on test bed
- Real temperature trace
- Queries request data at each sensor as Poisson process at the rate of 1 query/node/min
- Network is duty-cycled and can support at most 1 query/node/sec

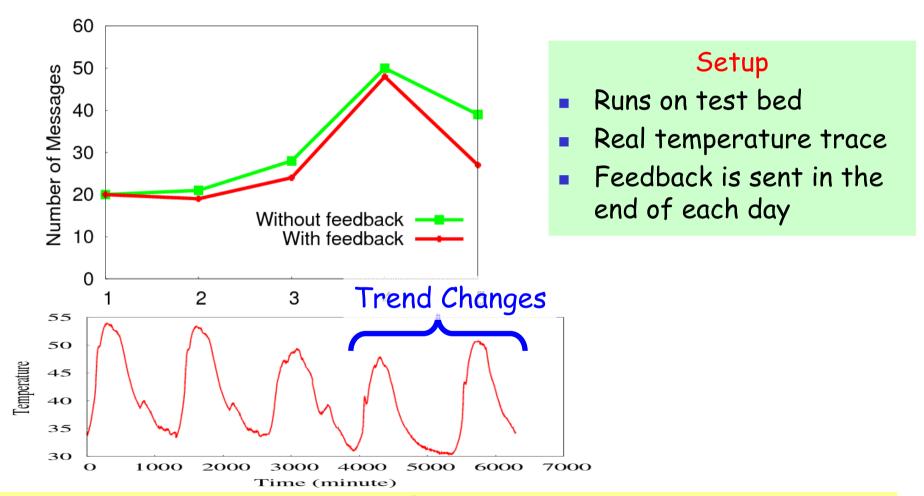


One proxy supports 100% more sensors than a sensor-centric system



Adapt to Data Dynamics

What the energy benefit of feedback-driven adaptation



Reduce communication by 30% compared to non-adaptive scheme

Summary

- PRESTO achieves low energy cost, low query latency and high accuracy by:
 - Splitting intelligence between the sensor and proxy using model-driven push and pull.
 - Using interpolation and prediction to maintain a precise proxy cache at low energy cost.
 - Feedback to update data and query model parameters.



Future Work

- Explore other models besides ARIMA that can be used to split complexity between the proxy and sensor.
- Extend to consider spatial correlation.
- Explore use of PRESTO in new sensor application domains.



The End

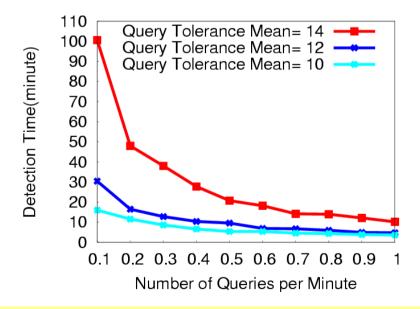
Thank You!

Questions?



Failure Detection

- Detection latencies in different query tolerances and query frequencies
- Setup
 - Runs on one Tmote
 - Using James Reserve data set
 - Queries arrive as Poisson arrival



Longest detection latency less than 2 hours

