

# Efficient Inference of Spatial and Temporal Conditions of Large Distributed Networks



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## Motivation

For a very large network with huge volume of time-varying data collected at thousands of distributed nodes, how to efficiently infer spatial and temporal conditions of the whole network?

## Our Proposed Approach

- Each node decides when to forward locally collected data to a centralized controller, subject to limited communication bandwidth
- The controller groups nodes into clusters dynamically based on supplied data, and derives a compressed representation of the whole network, which can infer conditions of unobserved nodes (Spatial Prediction)
- Use the compressed representation to predict future behavior of the network (Temporal Prediction)

## Spatial Predictions

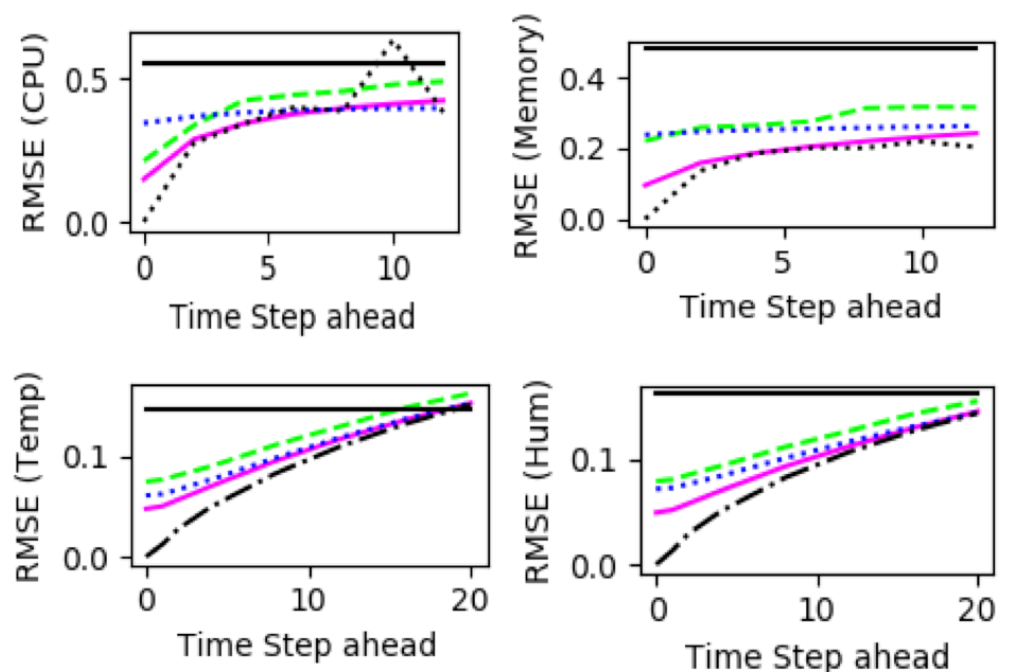
- Use mean values of B clusters of nodes to represent conditions of unobserved nodes
- Each node sends updated data to the controller when it expects a change of its cluster membership, inspired by Lyapunov optimization, subject to bandwidth constraint

## Spatial & Temporal Predictions

- **Best but non-scalable approach:** One forecasting model for each node with its root mean square error (RMSE) as lower bound for our proposed approach
- Our Proposed Approach: Use a forecasting model for each of B clusters and consider temporal correlation to identify the best forecasting model for a given node

— Proposed Approach      - - - Lower Bound  
- - - Minimum Distance      — Naive  
· · · Static clusters

(Note: 1 time step = 5 minutes ahead)

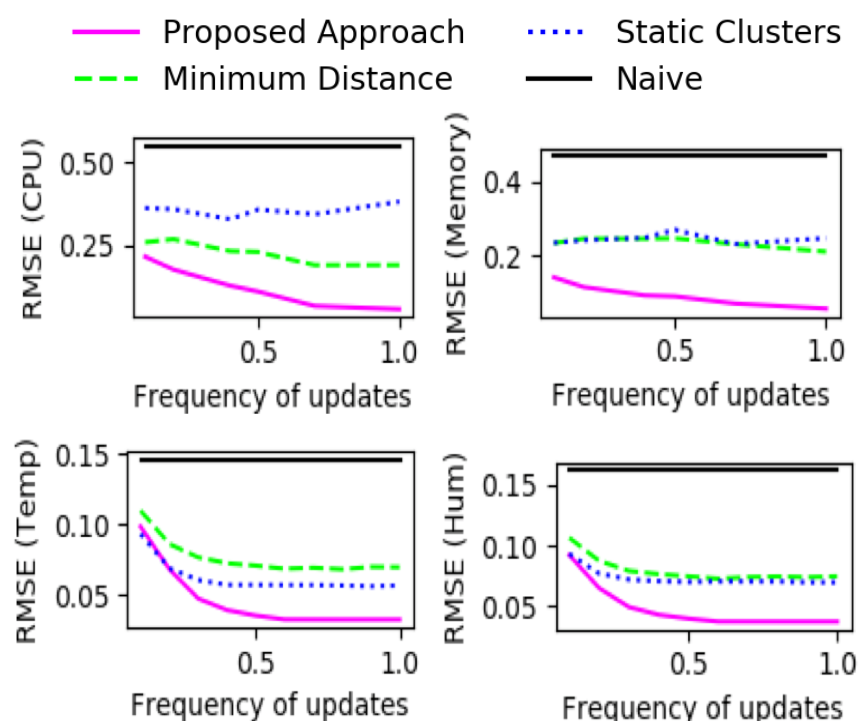


**Figure 2:** Spatial & temporal prediction at different steps ahead (freq. of updates = 0.3).  
 Top: CPU & memory usage of 100 machines, using 10 forecasting models ( $B = 10$ ).  
 Bottom: 47 temperature & humidity sensors using 3 forecasting models ( $B = 3$ ).

### Comparison with Two Other Approaches:

1. Spatial predictions using data from B randomly chosen nodes. An unobserved node is treated as one of the B nodes based on minimum distance (Minimum Distance)
2. Predictions by data averaged across nodes of static clusters (Static Clusters)

Note: Our proposed approach is similar to Approach 2, but with dynamic clustering.



**Figure 1:** Spatial prediction.  
 Top: CPU & memory usage of 100 machines,  $B = 10$ .  
 Bottom: temperature & humidity of 47 sensors,  $B = 3$ .