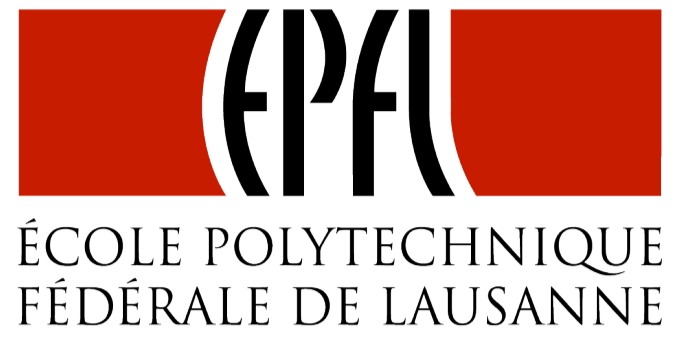


## OptiMoS: Optimal Sensing for Mobile Sensors

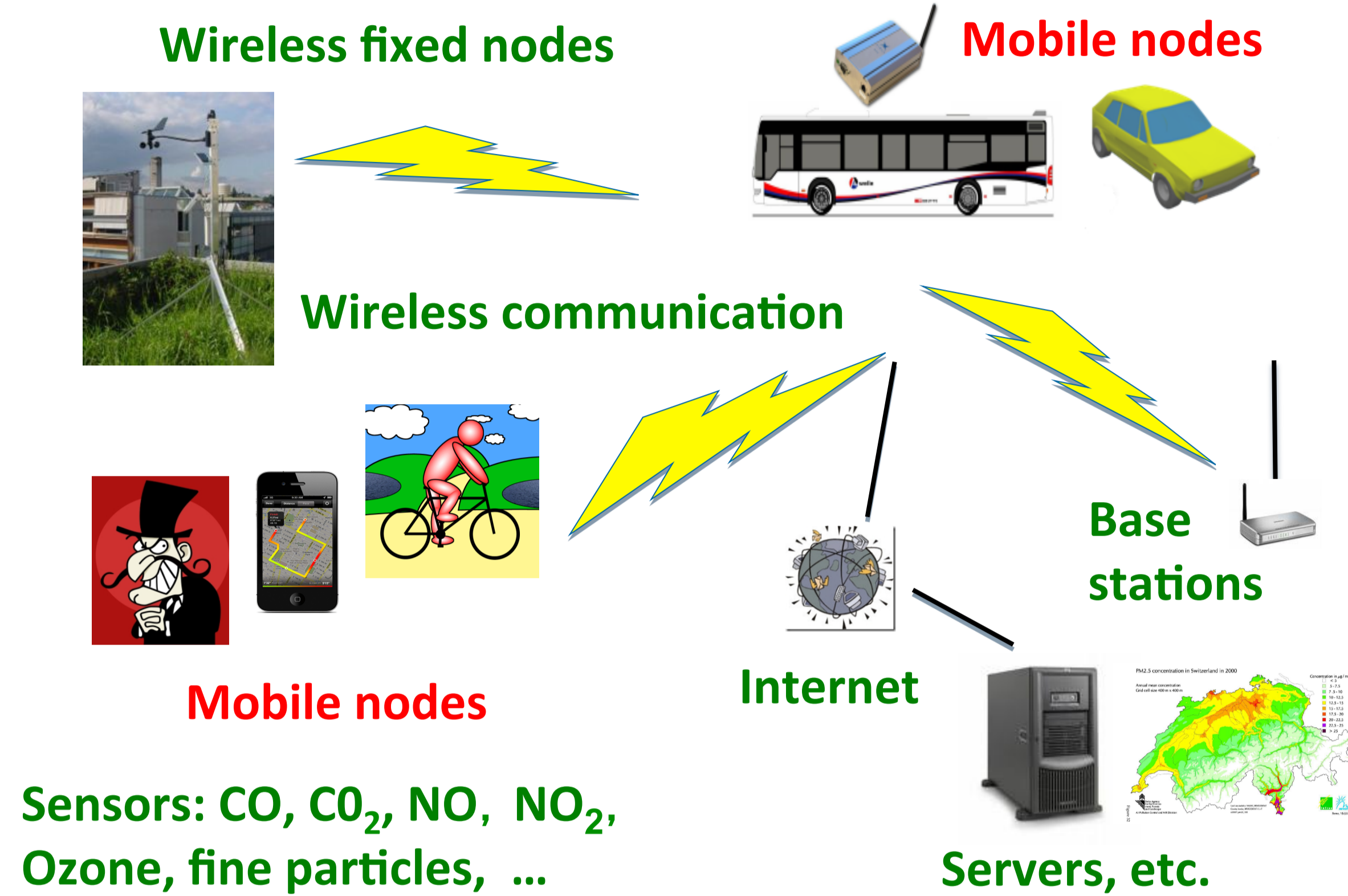


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<sup>1</sup> LSIR-IC-EPFL    <sup>2</sup> Nokia Research Center - Lausanne

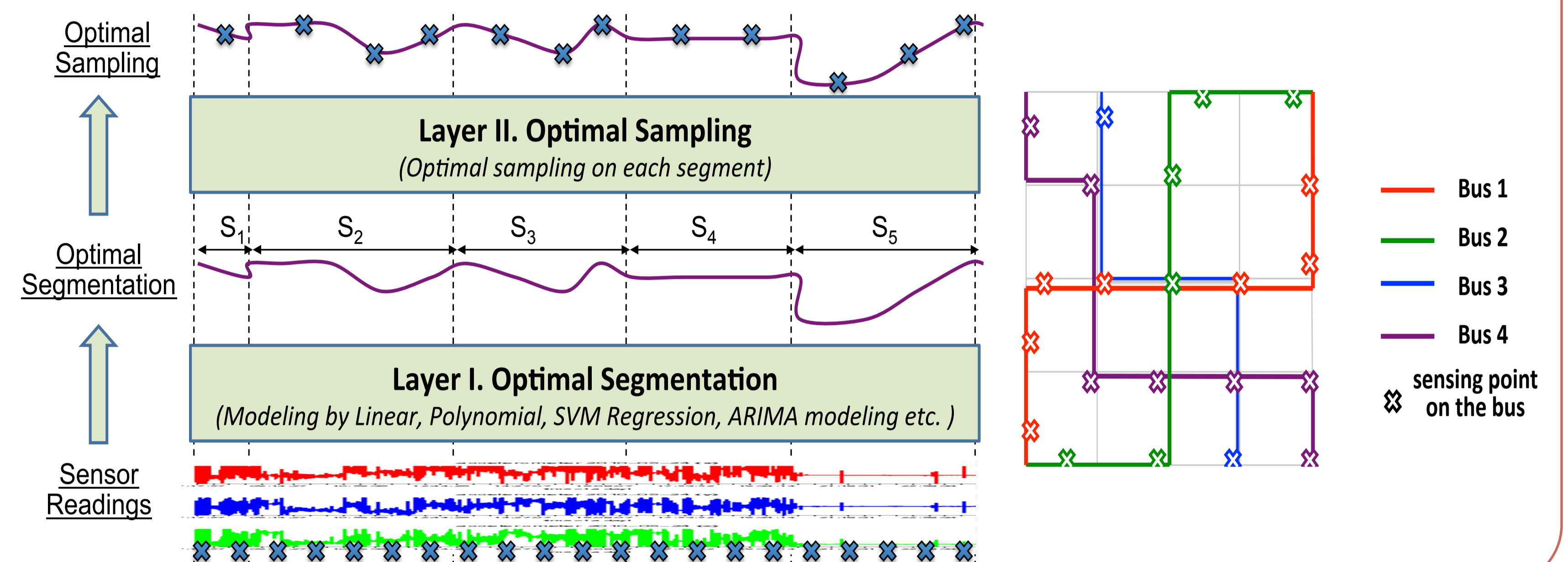


**OpenSense: An open environmental sensing platform with massive, heterogeneous, mobile and increasingly miniaturized sensors.**



**OptiMoS: An optimal mobile sensing for achieving appropriate tradeoff between "sensor coverage maximization" and "energy cost (sampling) minimization".**

- ❖ Two-Tier Sensing Platforms: Segmentation and Sampling
- ❖ Optimal sensing for one bus line (a single mobile node), or multiple bus lines
- ❖ Optimal co-sensing among different sensors, e.g., CO, CO<sub>2</sub>, NO<sub>2</sub> ...



## Model-Driven Segmentation of Mobile Sensing

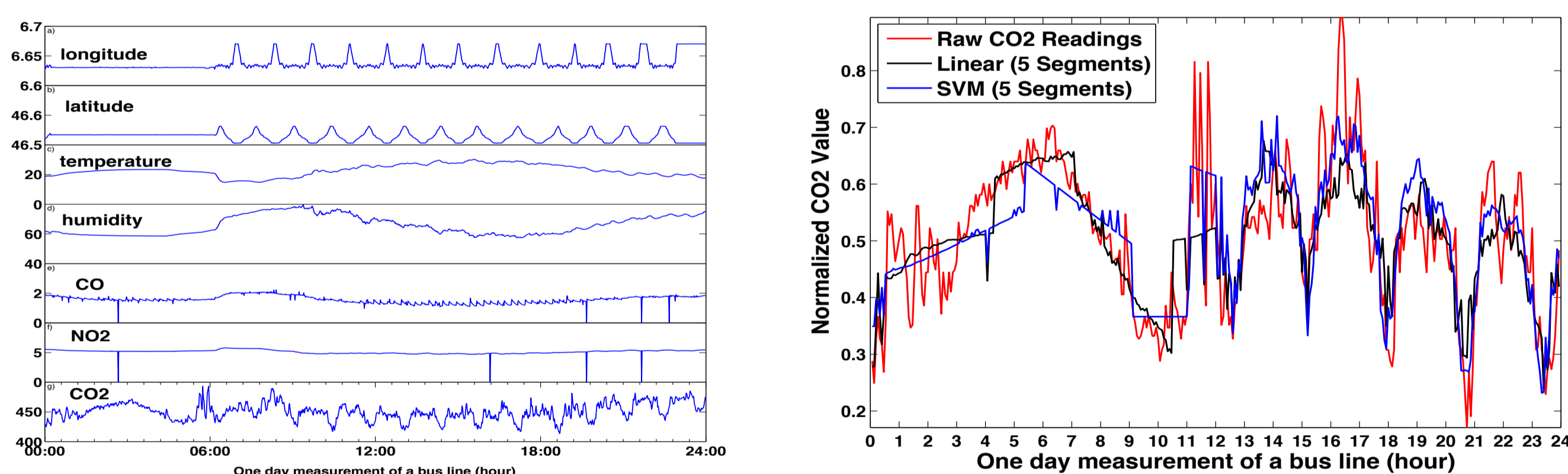
### Problem Statement of Segmentation

- ❖ **Initial Sensor Readings**
  - Sensor reading records
  - Each record with time, location, measurements
- ❖ **Data-driven Modeling**
  - Various models: linear, SVM regression, etc.
  - Model errors RSS – Residual Sum of Square
- ❖ **Optimal Segmentation**

$$\mathcal{R} = \{R_1, \dots, R_N\}$$

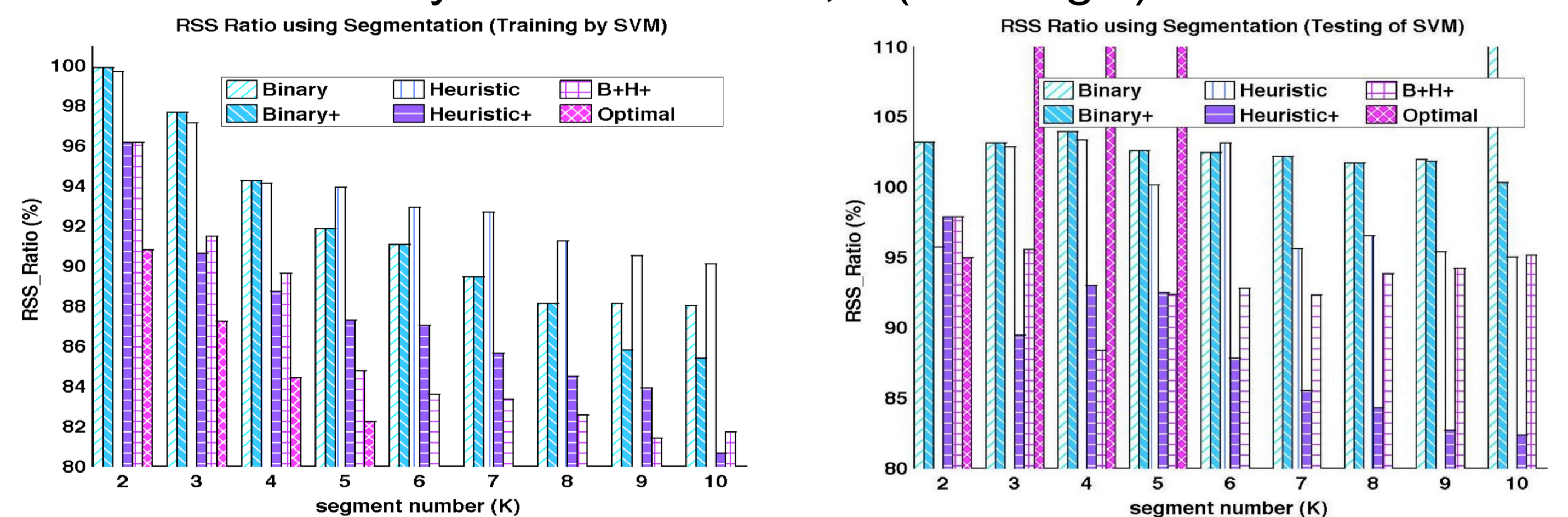
$$R_i = \langle t, l, x_1, \dots, x_m \rangle$$

$$\underset{K, d_1, d_2, \dots, d_{K-1}}{\operatorname{argmin}} \sum_{i=1}^K \operatorname{RSS}(\mathcal{M}(\{R_{d_{i-1}}, \dots, R_{d_i}\}))$$



### Segmentation Strategies

- ❖ **Optimal Segmentation**
  - Dynamic programming, expensive O(K\*N<sup>2</sup>), over-fitting
- ❖ **Top-down Binary Segmentation**
  - Binary: O(K\*logN)
  - Binary+: better strategy in finding division segment
- ❖ **Error-based Heuristic Segmentation**
  - Heuristic: division by absolute errors
  - Heuristic+: division by relative errors
- ❖ **Near-Optimal Segmentation**
  - B+H+: Binary+ with Heuristic+, O(K\*N\*logN)



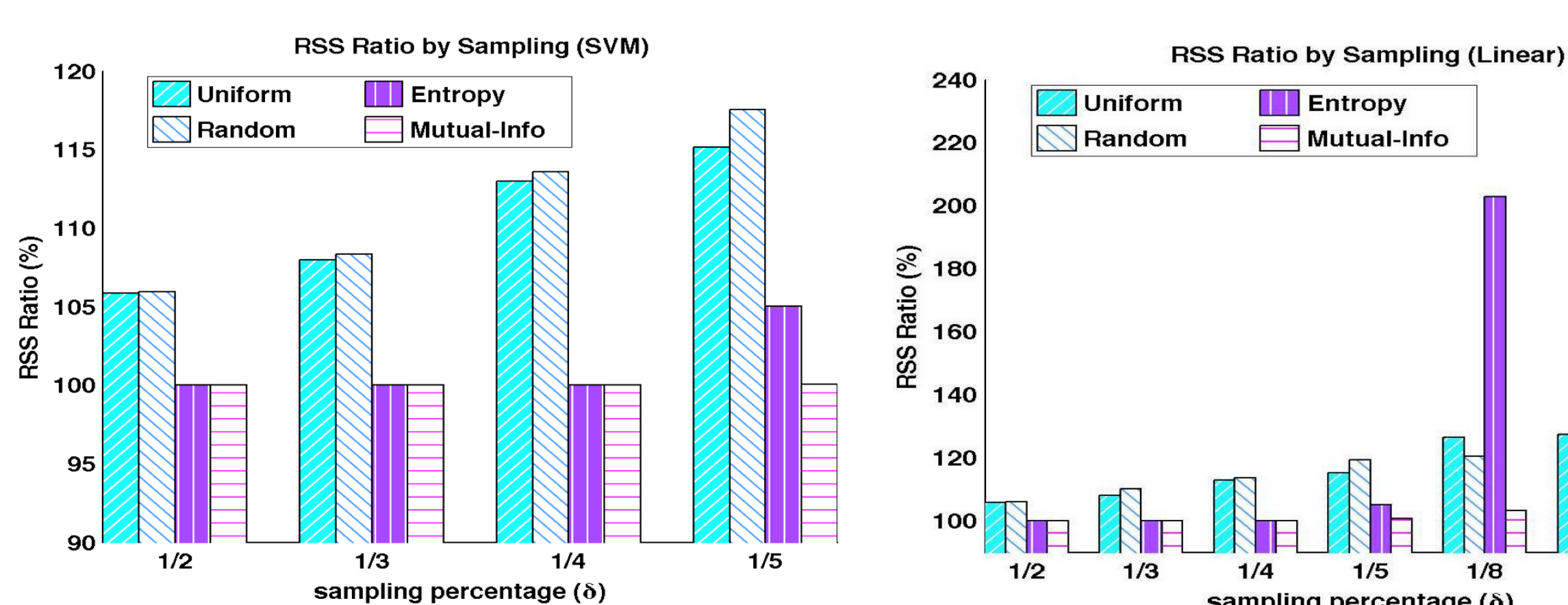
## Near-Optimal Sampling for Individual Segments

### Sampling Strategies

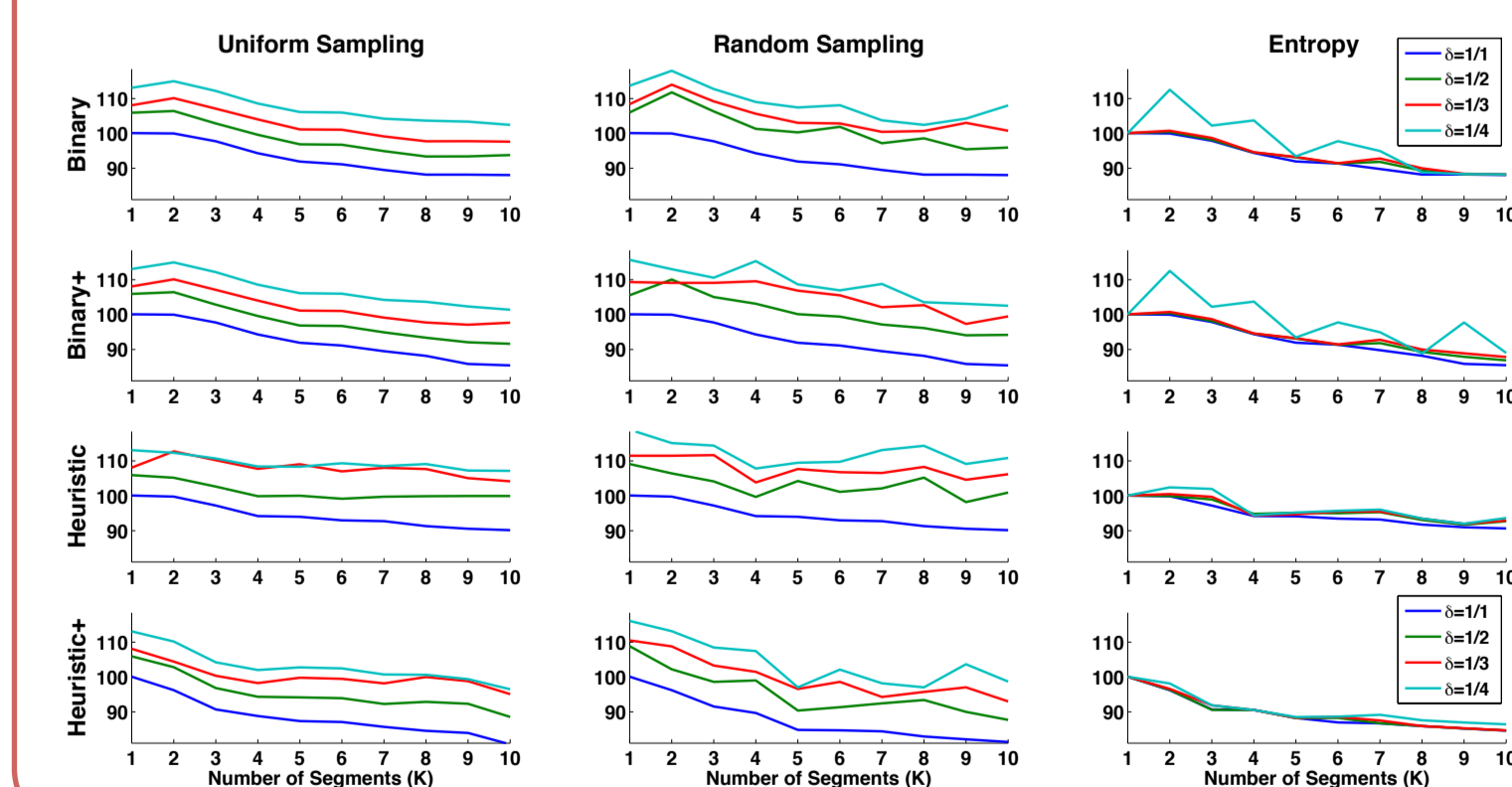
- ❖ **Optimal Sampling**
  - NP-hard problem
- ❖ **Distribution-based Sampling**
  - Uniform: regular duty cycle readings
  - Random: irregular duty cycle readings
- ❖ **Entropy (Error) based sampling**
  - Selecting points with top entropy
- ❖ **Mutual Information based sampling**
  - Remove information redundancy
  - Recalculate entropy after each selection/sampling

$$\underset{\mathcal{R}_{sub}}{\operatorname{argmin}} \mathcal{L}(\mathcal{R}, \mathcal{R}_{sub}) \quad \text{s.t.} \quad |\mathcal{R}_{sub}|/|\mathcal{R}| \leq \delta$$

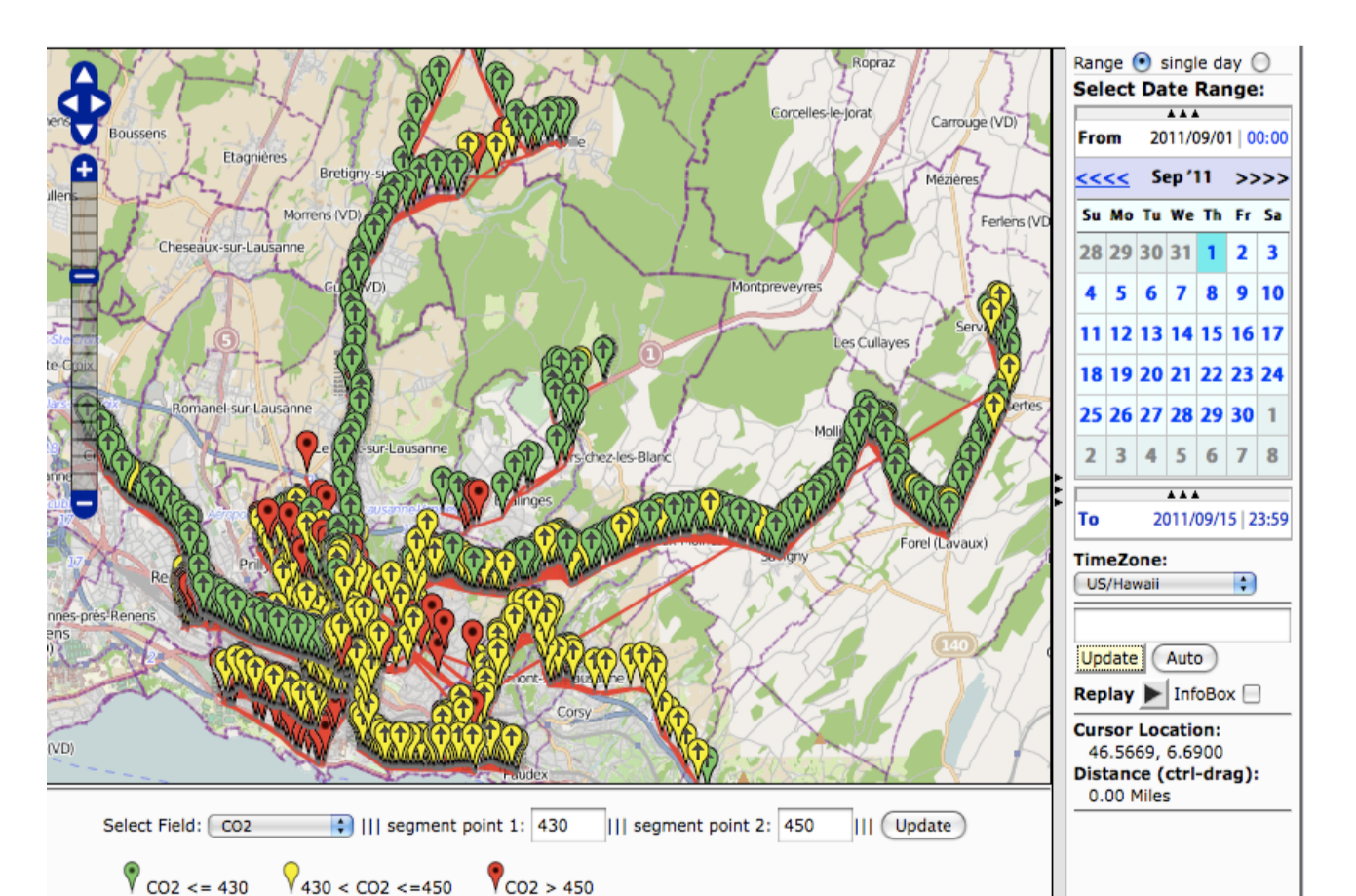
$$\underset{\mathcal{R}_{sub}}{\operatorname{argmin}} |\mathcal{R}_{sub}|/|\mathcal{R}| \quad \text{s.t.} \quad \mathcal{L}(\mathcal{R}, \mathcal{R}_{sub}) \geq \epsilon$$



### Two-tier Combinations



**Visualization: create pollution map by model prediction from limited measurements.**



### Integration with GSN for real-time data prediction

