



# Estimating Power Loads from Partial Appliance States

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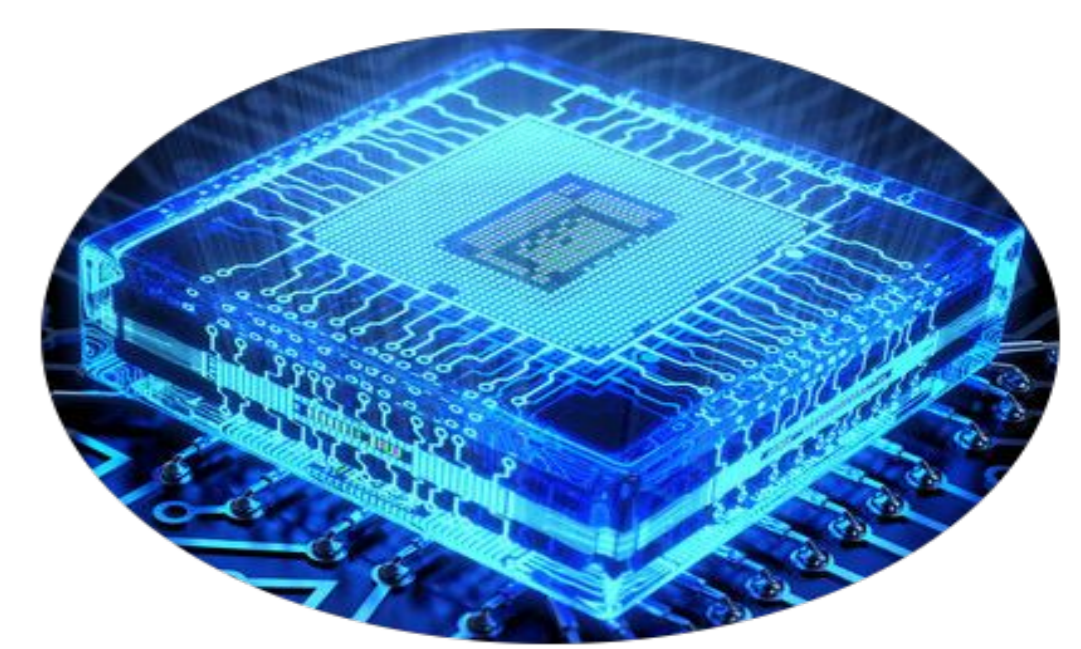
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# Estimating Power Loads from Partial Appliance States

Nicolas Roux (nicolas.roux@inria.fr), Baptiste Vrigneau, Olivier Sentieys

## General scope

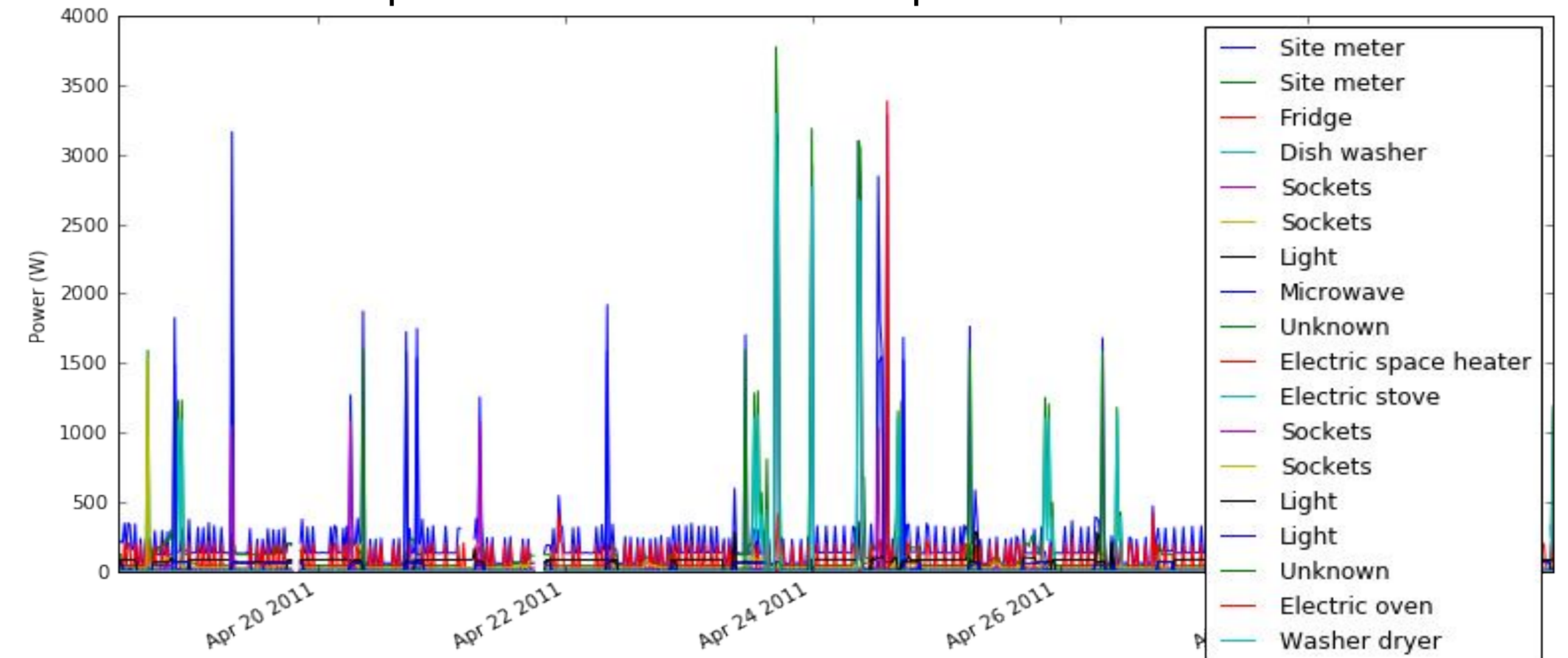
Knowing the plug-level power consumption of each appliance in a building can lead to significant savings in energy consumption, as well as giving a deep insight of a building power consumption. A critical information is the knowledge of the characteristic power consumptions of an appliance. The problem can be stated as an optimization problem as follows :

$$\min_{w_{i,j}} \left\| x_{tot}(t) - \sum_{i=1}^N \sum_{j=1}^{M_i} w_{i,j} \times s_{i,j}(t) \right\|_d \quad (1)$$

- $w_{i,j}$  the steady state power value  $j$  for appliance  $i$
- $x_{tot}$  the main aggregated power of all appliances
- $N$  the total number of monitored appliances
- $M_i$  the states count for appliance  $i$
- $s_{i,j}$  a boolean which indicates if appliance  $i$  is in state  $j$
- $d$  a given norm.

## Non-Intrusive Load Monitoring (NILM)

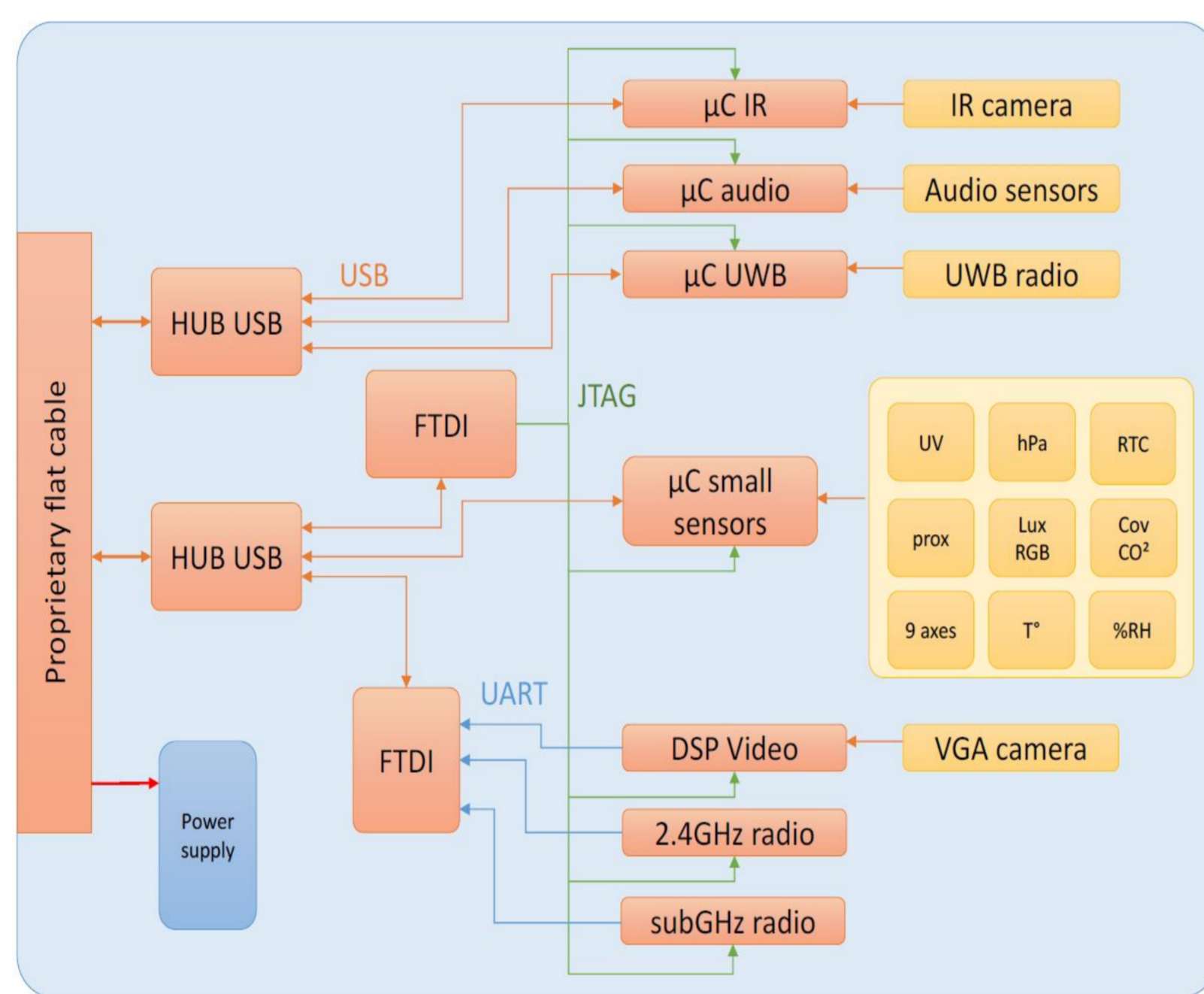
An example of NILM on the REDD public dataset House 1



Main power of a building is disaggregated to isolate single appliance power consumptions. This ensures a better understanding of the energy consumption in a building. Most of the current NILM methods have an *a priori* knowledge of the individual signatures or are trained on single appliances.

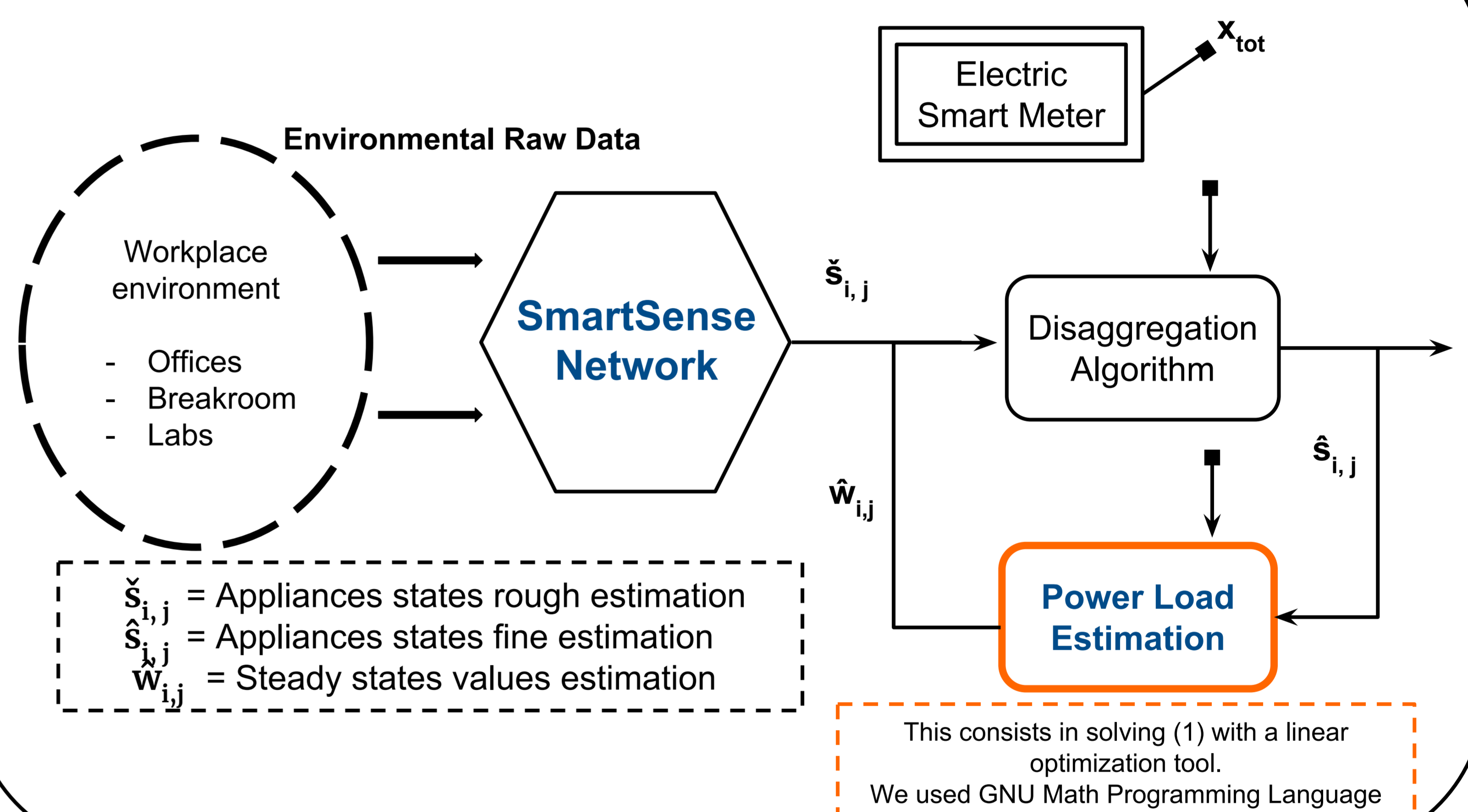
## SmartSense

- We believe that additional info can help in improving disaggregation algorithms
- Basic computations made directly on sensor node
- Ensures low network usage and user privacy
- Environment is an electronic lab
- 156 nodes are deployed
- More info and data download at [smartsense.inria.fr](http://smartsense.inria.fr)



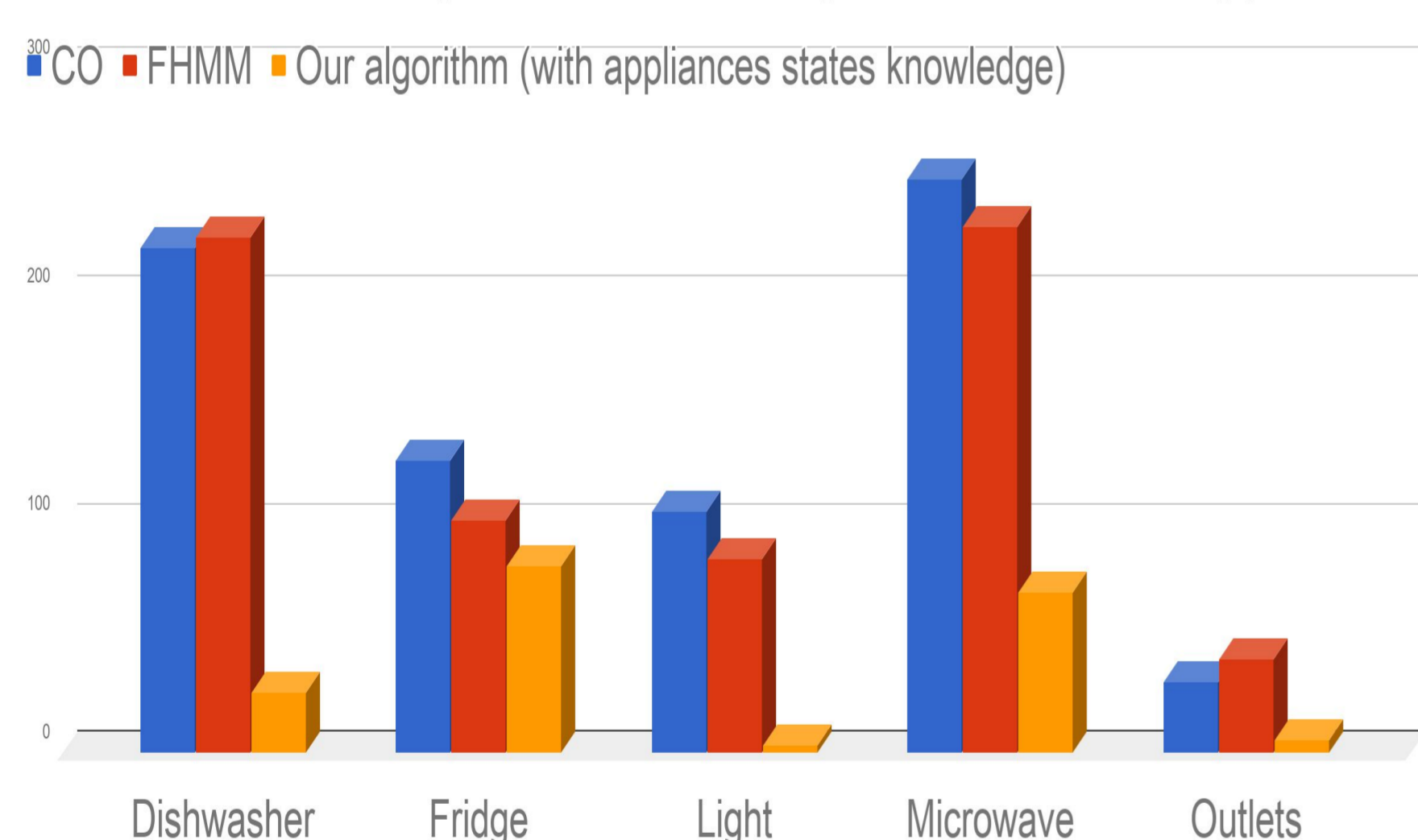
Block-diagram of a SmartSense Node

## Dataflow

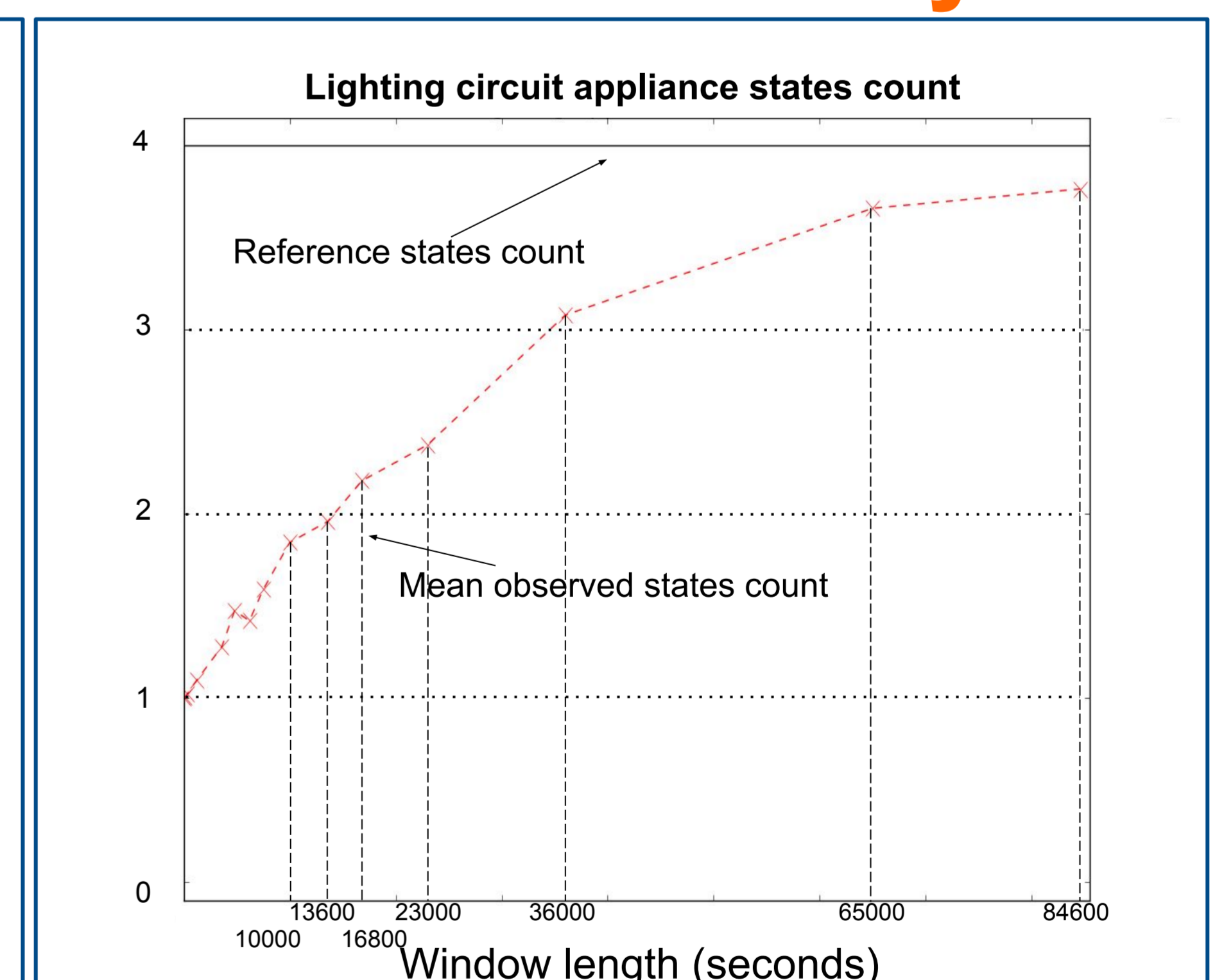
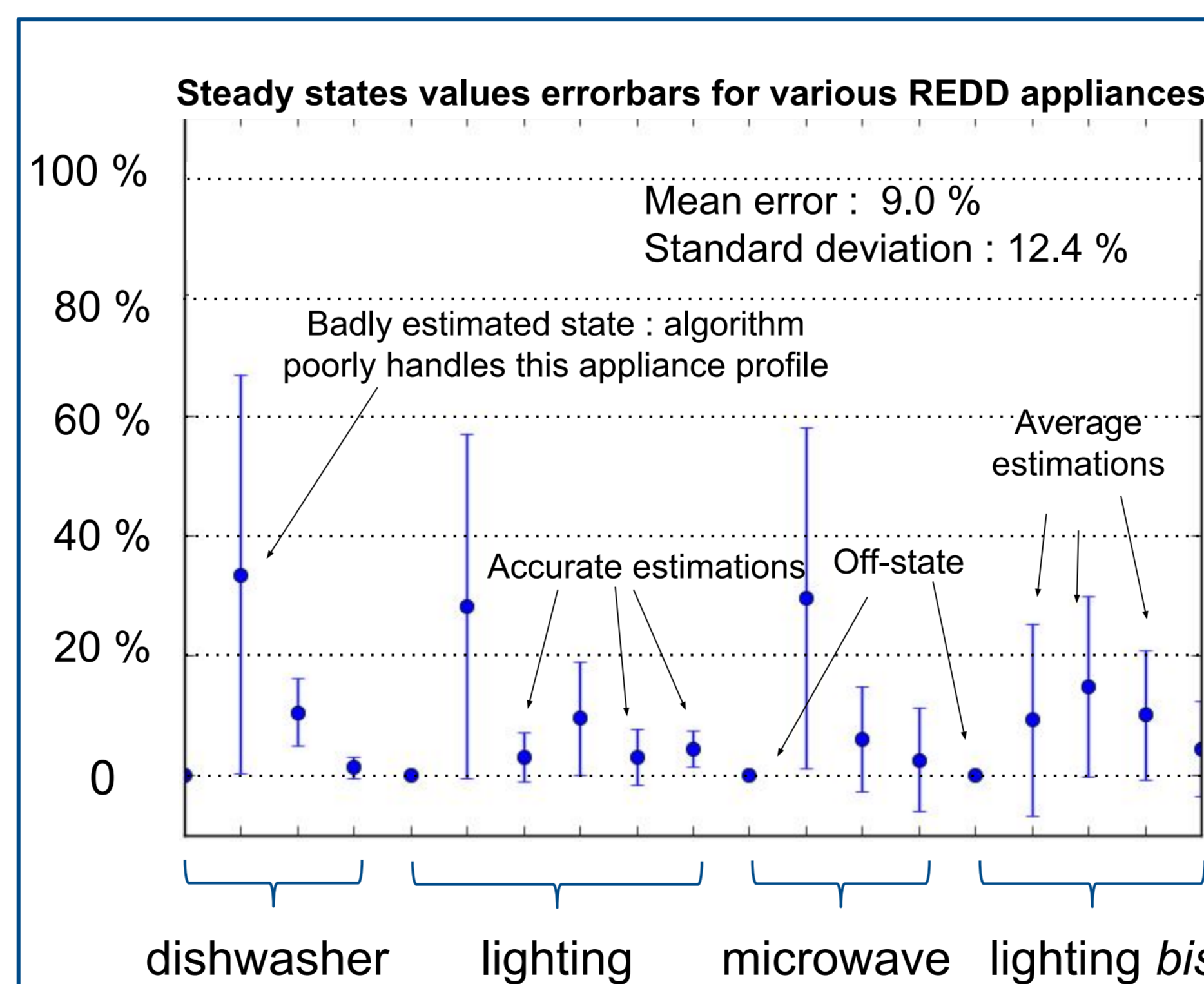


## Comparison with "blind" algorithms

RMS error between predicted and true power for REDD 1 appliances



## States detection and estimation accuracy



## Conclusion

- An estimation algorithm based on environmental sensing has been proposed.
- SmartSense, a sensor network with a wide range of sensors is introduced.
- Promising results are displayed in terms of complexity and prediction error, but the algorithm showed significant variations according to the nature of the appliance power time series.

## Future work

- Smartsense deployment and data collection are imminent.
- Smartsense is expected to yield accurate environmental data but inaccurate appliance states matrix.
- Next challenges to address are the algorithm robustness to this kind of errors, and association between NILM algorithms and environmental data instead of electrical data.

