

Energy-Efficient Activity Recognition using Prediction

ISWC 2012 – Newcastle, England

Dawud Gordon, Jürgen Czerny, Takashi Miyaki, Michael Beigl
Karlsruhe Institute of Technology (KIT), TecO





Motivation

■ What happens as the number of devices increases?

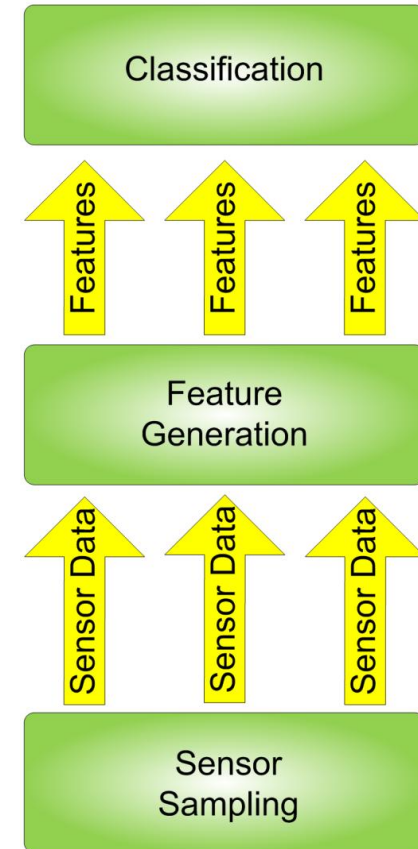


Use Prediction to Save Energy

Context prediction: forecast future contexts!

- Classify activity
- Predict probable future activities
- Identify sensor subset for differentiation
- Activate sensor set
- Repeat

- Save energy!!



What is there to research?!

Research Issues

- How to select sensors for to be turned off?
 - How much information is delivered to

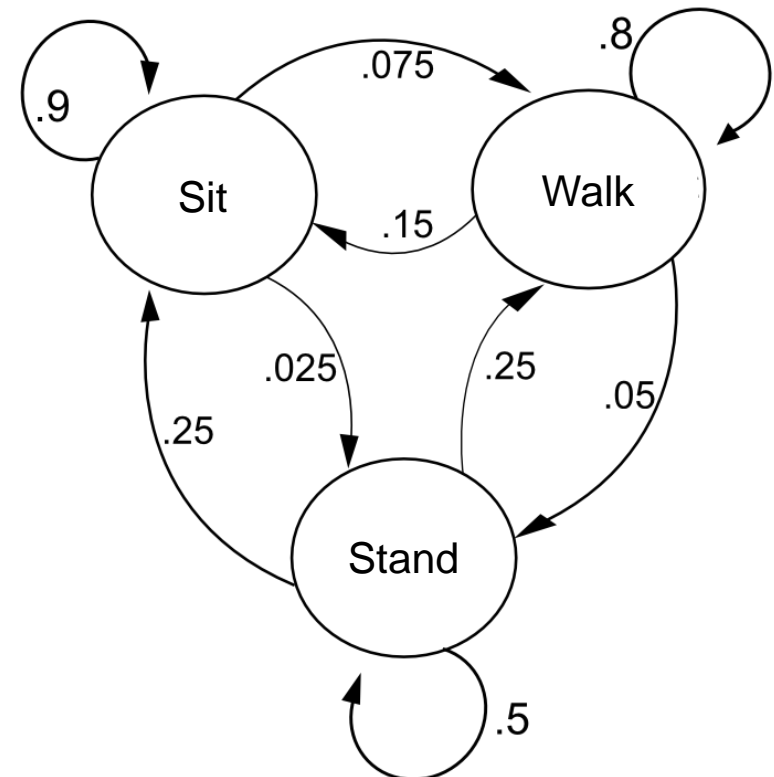
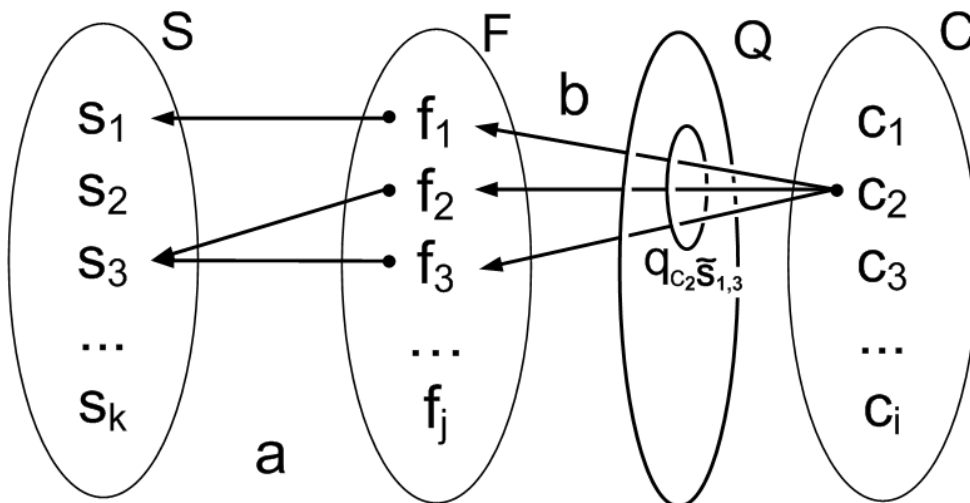
- How many classes to predict (ρ)?
 - 1 class prediction?
 - All class prediction?

- And what is the acceptable cost (λ)?

- How does performance depend on scenario?
 - How predictable is the subject (κ)?

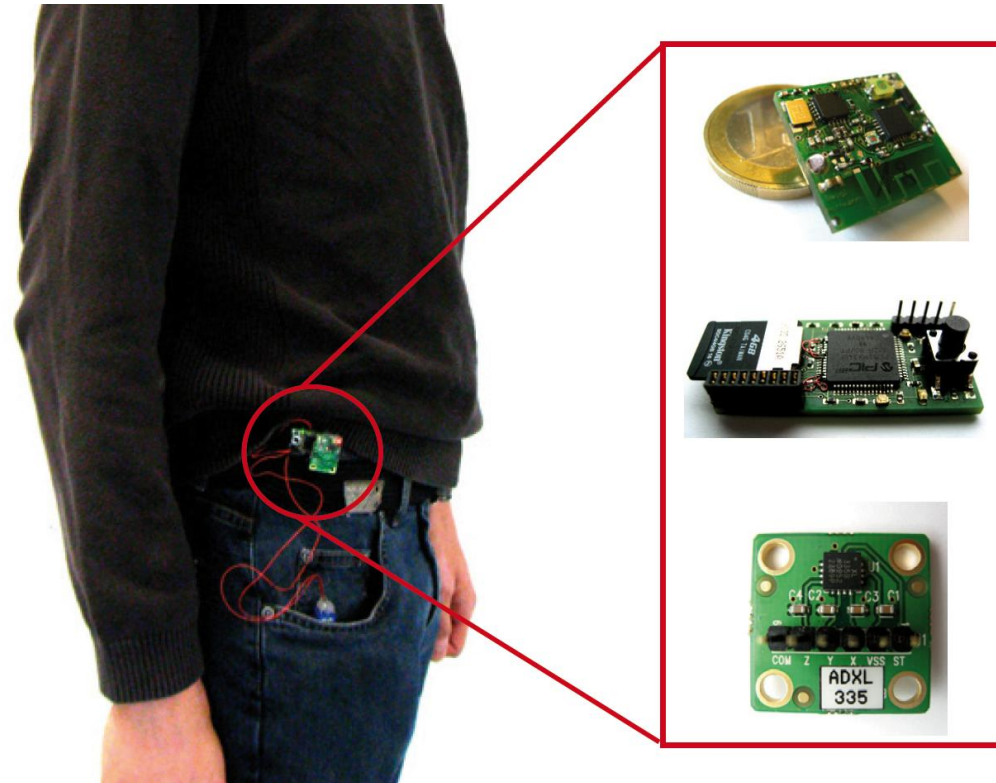
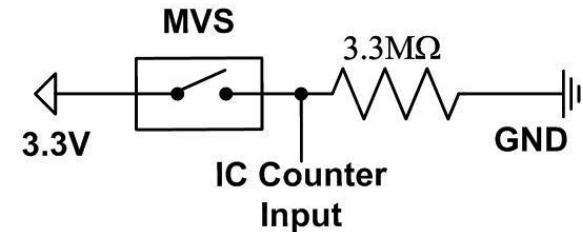
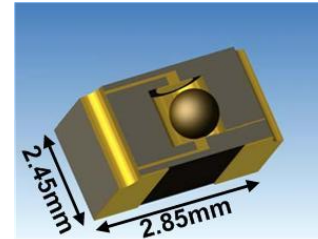
System Description

- Activity recognition algorithms
 - K-Nearest Neighbors
 - Hidden Markov Model
- Prediction algorithm
 - Markov chain
- Method for mapping activities to sensors
 - Selecting S without violating λ



Activity Recognition Data Set

- 5 Subjects
- 8 Activities
- 60 Hz sampling
 - Vibration
 - Acceleration
 - Light
 - Temperature
- 142 Minutes of data
- Dataset available at:
www.teco.kit.edu/~gordon/MVS/



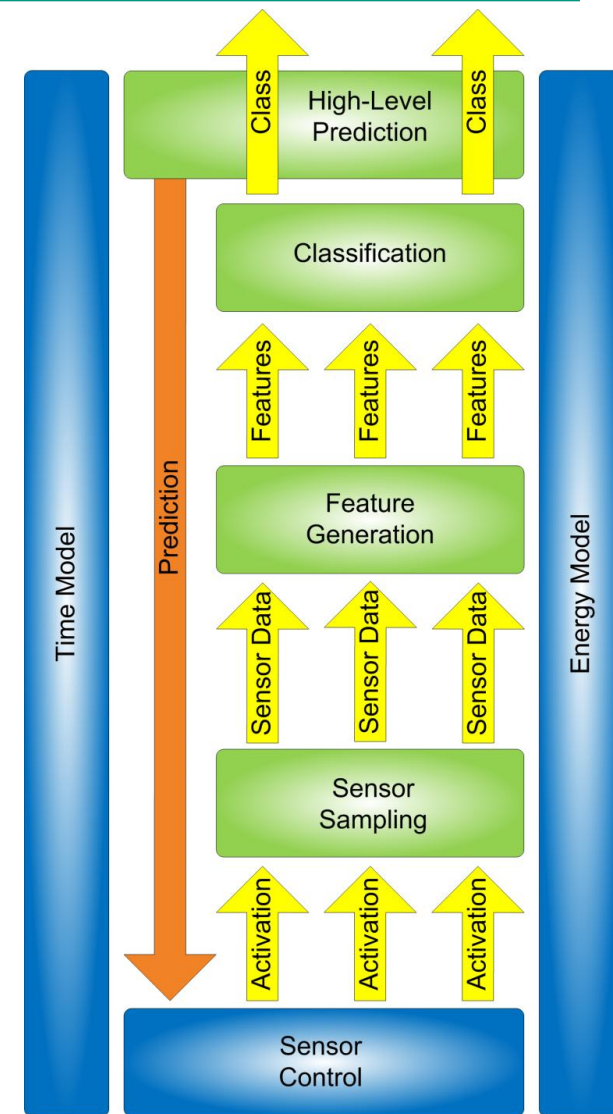
Gordon et al. A Novel Micro-Vibration Sensor for Activity Recognition: Potential and Limitations, ISWC'10

Simulation Environment

- Generate synthetic data sets using generative model (HMM)
 - Vary predictability (κ) (0.125:0.875 step 0.125)
 - Train: 2/5 train, 1/5 test, 2/5 eval

- Simulate performance
 - Loss parameter (λ) (0:100 step 10)
 - Predicted states (ρ) (1:8 step 1)

- Evaluate
 - Evaluate base case (ALL ON)
 - Classifier accuracy (loss)
 - Energy consumption (savings)



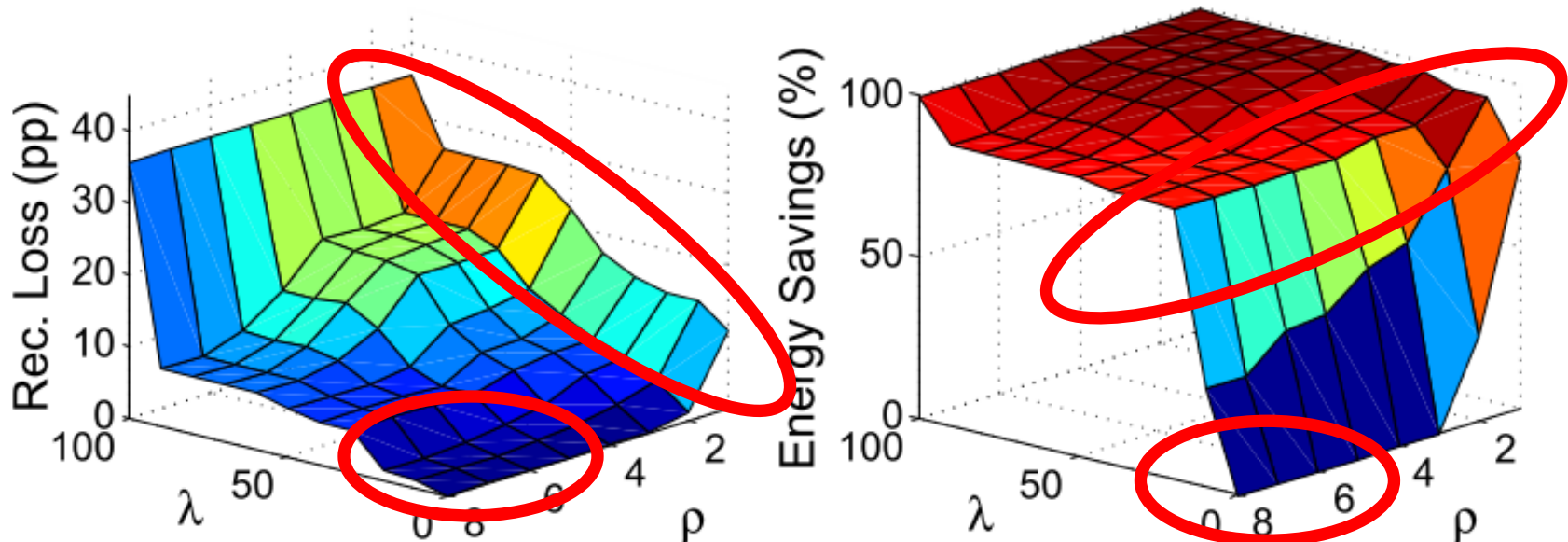
Lots of Results!

Multi-dimensional results matrix!

- $[\kappa, \lambda, \rho, \text{classifier}]$
- A matrix for acc., prec., rec., f-meas., energy
- Confusion matrix 2 more dimensions!

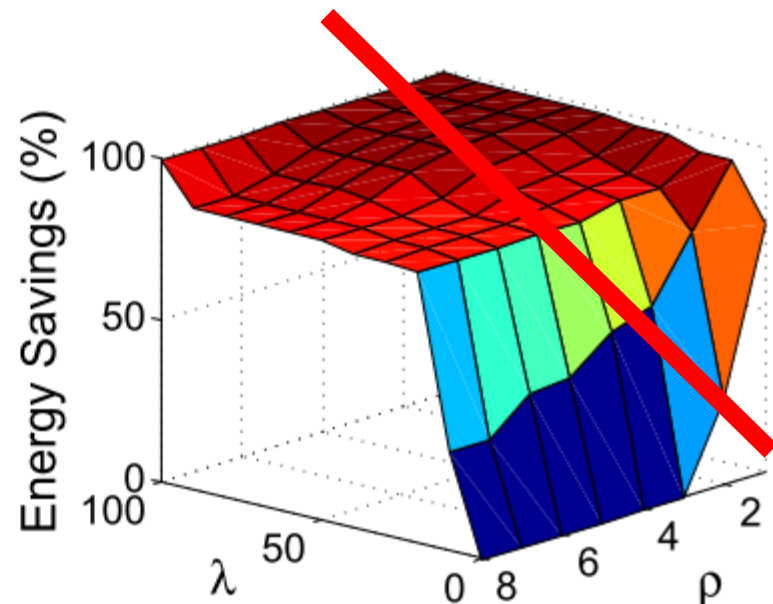
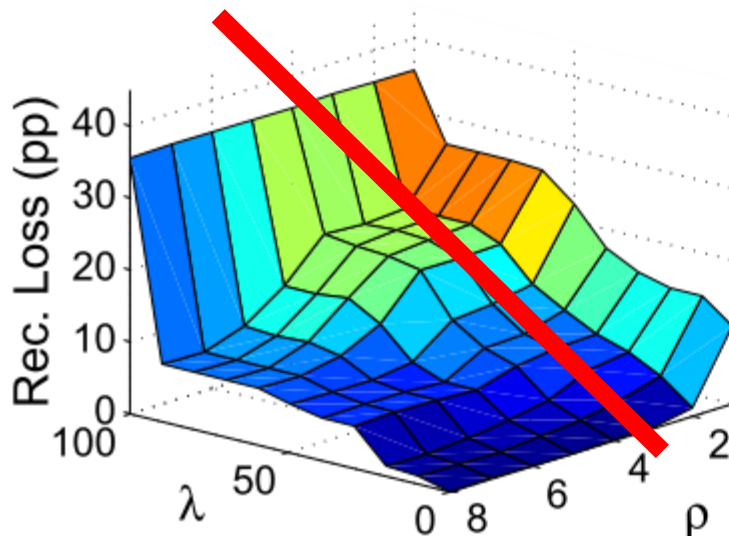
This work:

- Accuracy (loss) and energy (savings)



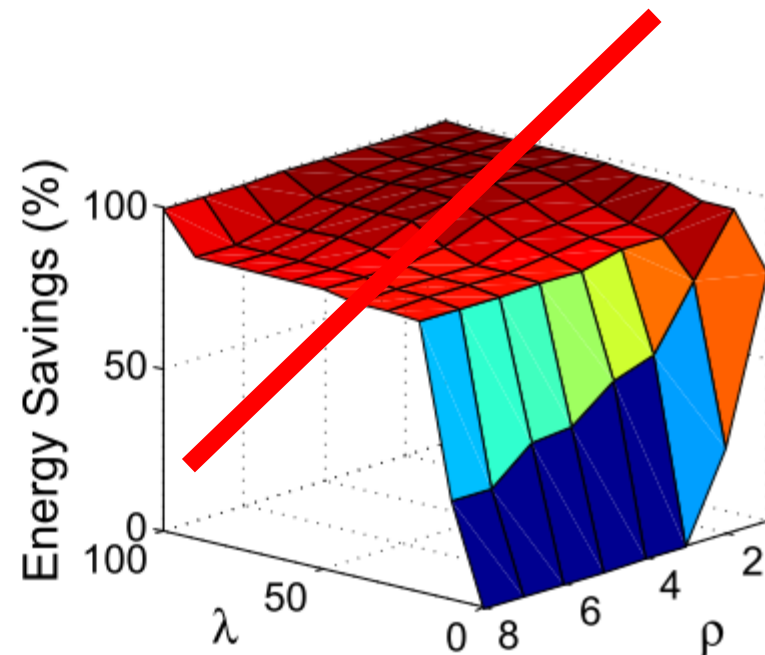
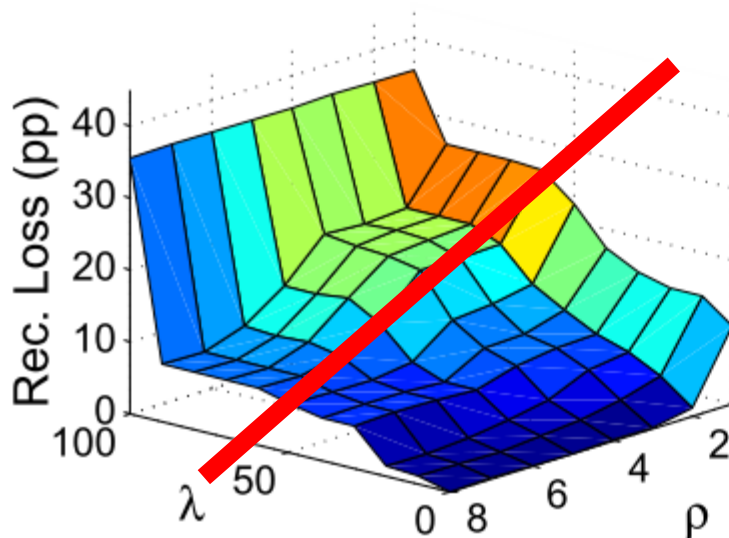
Loss Parameter λ

- Defines amount of recognition to trade for energy
- Like a **wager!**
- Monotonic increase in for fixed ρ , κ
 - Loss of recognition
 - Energy savings



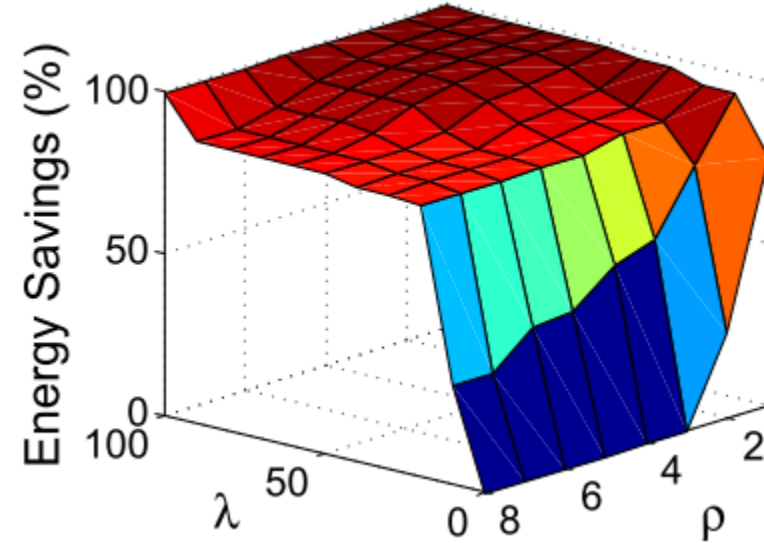
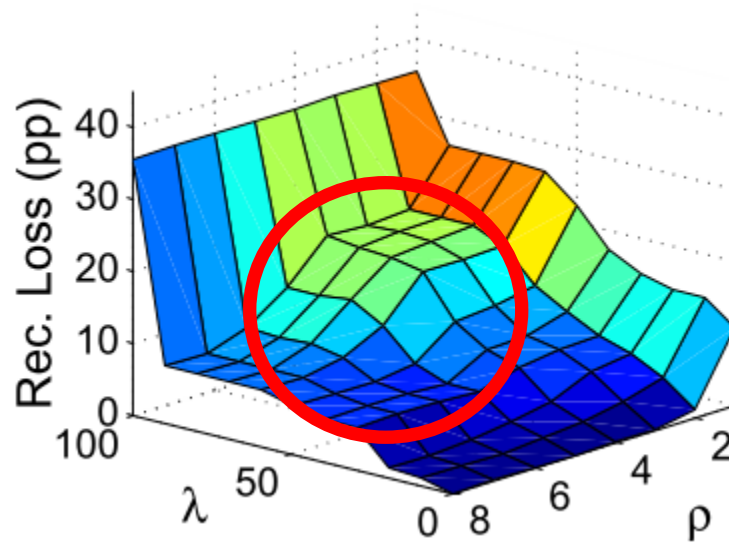
Number of Predicted States ρ

- Defines **amount of risk!**
- High risk – high payout
- Monotonic **decrease** in for fixed λ, κ
 - Loss of recognition
 - Energy savings

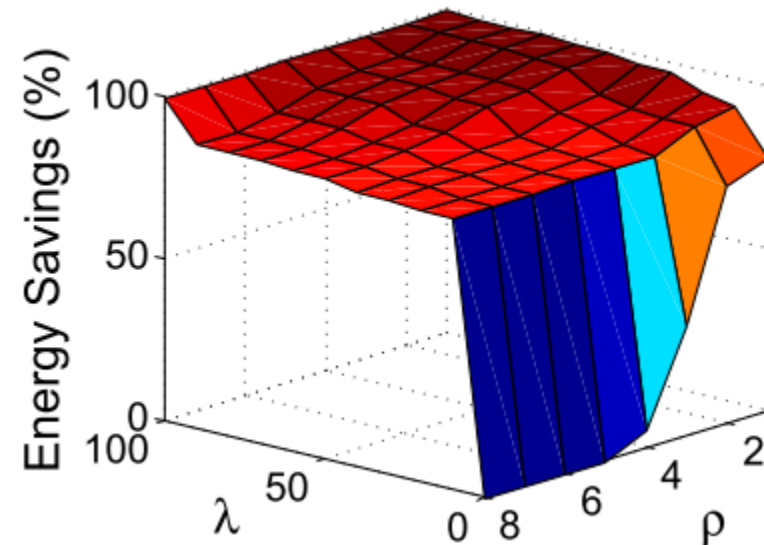
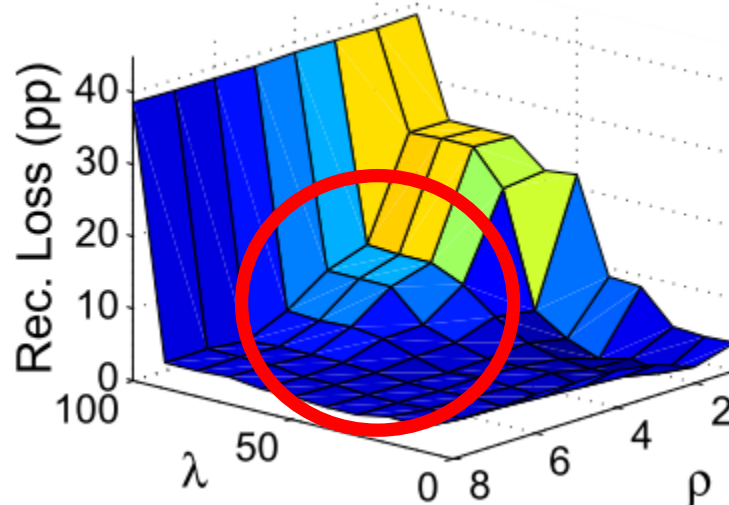


Effects of Predictability κ

$\kappa = .125$



$\kappa = .875$



Conclusion

- Novel approach using prediction
- Saves energy, e.g. 84% for 1.2pp
- Results heavily dependent on
 - Sensors used
 - Activities recognized
 - Predictability of scenario
- Tools for tuning desired trade-off supplied!
 - κ can be obtained with predictor and scenario
 - λ taken from the application
 - ρ adjusted to find the desired give and take

Thank you and Questions!

Thank you and Questions!

- So much more data!
- Effects of sensor control on prediction
- OPPORTUNITY data set

HELP: Special Issue CfP?!?