Activity Recognition in a home

Patrick, Dong, Stephen

Activity Recognition

 Recognizing human activities e.g. home occupancy, sleeping, cooking, bathing







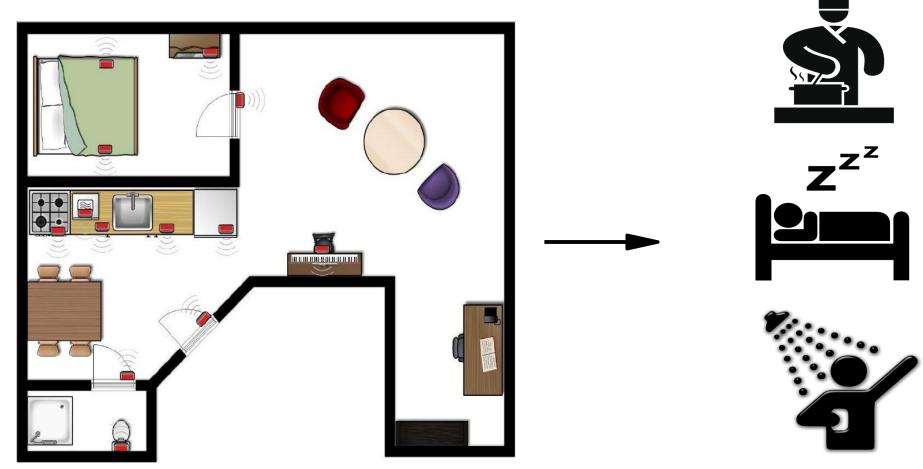
Nest

Ulo

Belkin Smart Switch

Problem definition

 Infer occupant activities using sensors home deployed sensors

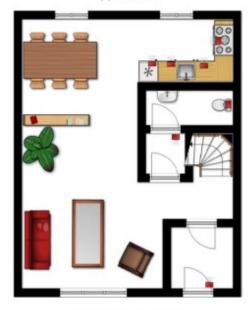


Evaluation

- 2 apartments1 home
- Each single occupant
- Duration14 25 days



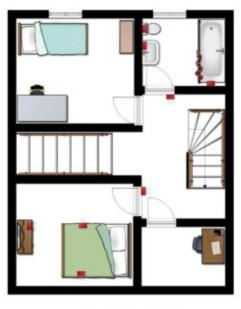
(a) House A



(c) House C, First floor



(b) House B



(d) House C, Second floor

Dataset

- Binary sensor data:
 - Reed switches: doors or cupboards open or closed
 - Pressure mats: lying on bed or couch
 - Passive infrared: Detect motion in specific area
 - Float sensors: toilet being flushed
- Labels:
 - Handwritten diary
 - Bluetooth diary

Related work

- Van Kasteren, T. L. M., Gwenn Englebienne, and Ben JA Kröse. "Human activity recognition from wireless sensor network data: Benchmark and software." *Activity* recognition in pervasive intelligent environments. Atlantis Press, 2011. 165-186
- Tapia, Emmanuel Munguia, Stephen S. Intille, and Kent Larson. Activity recognition in the home using simple and ubiquitous sensors. Springer Berlin Heidelberg, 2004.

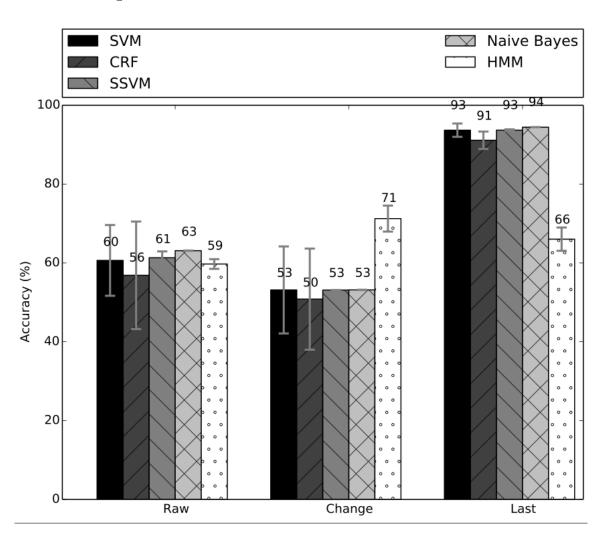
Methods

- Naïve Bayes
- SVM
- HMM
 - Generative probabilistic model
- CRF
 - Discriminative probabilistic model
- Structured SVM
 - Allows training of a classifier for general structured output labels

Evaluation Design

- Feature representations
 - Raw
 - Change
 - Last
- 5 fold cross validation
- Divide the dataset into smaller subsequences of ~2 hours

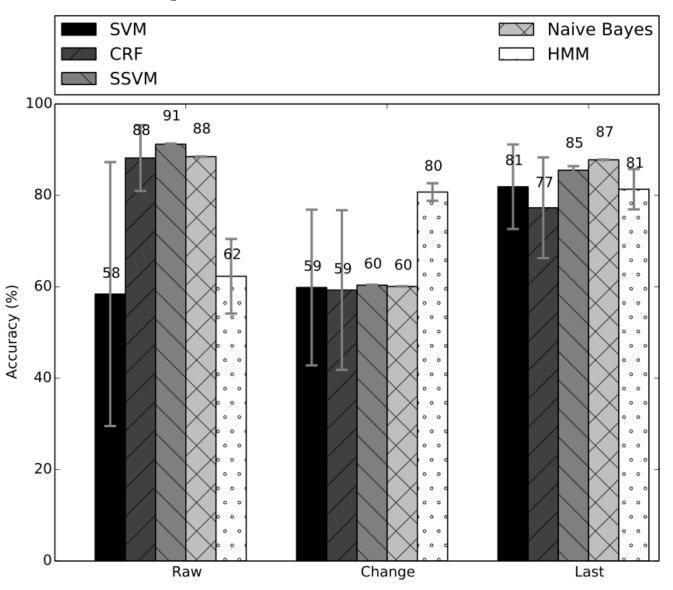
Comparison: House A



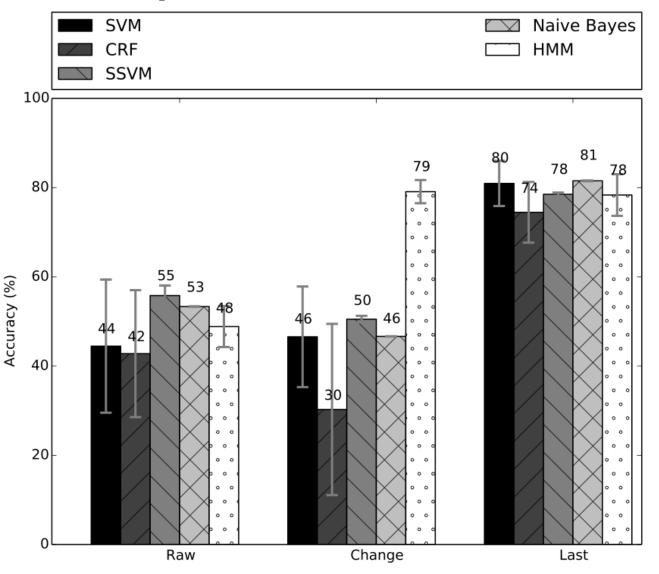
House A Metrics

		Precision	Recall	F-Measure	Accuracy
Model	Feature				20
HMM	Change	36.9	35.2	31.8	72.0 ± 3.3
	Raw	30.8	13.8	12.9	58.0 ± 1.2
	Last	21.8	19.3	15.2	66.6 ± 3.0
SVM	Change	34.7	11.4	9.5	53.1 ± 11.0
	Raw	33.2	17.0	17.4	60.6 ± 9.0
	Last	26.7	27.9	27.3	93.7 ± 1.7
SSVM	Change	13.9	10.3	7.5	53.1 ± 0.0
	Raw	23.0	15.2	15.3	61.3 ± 1.6
	Last	30.3	29.2	28.9	93.7 ± 0.2
CRF	Change	3.1	10.0	4.8	31.3 ± 0.0
	Raw	17.9	15.7	14.9	62.1 ± 0.0
	Last	19.4	19.3	14.5	65.9 ± 0.0
$_{\rm HMM}$	Change	36.9	35.2	31.8	72.0 ± 3.3
	Raw	30.8	13.8	12.9	58.0 ± 1.2
	Last	21.8	19.3	15.2	66.6 ± 3.0
NB	Change	38.1	11.6	9.9	53.1 ± 0.0
	Raw	52.7	17.3	17.0	63.0 ± 0.0
×	Last	28.3	28.2	27.7	94.5 ± 0.0

Comparison: House B



Comparison: House C



Summary

- Naive Bayes consistently outperforms other methods
- Structured SVM also performs well in sensors based activity recognition.