

Activity Recognition in a home

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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 apartments
1 home
- Each single
occupant
- Duration
14 – 25 days



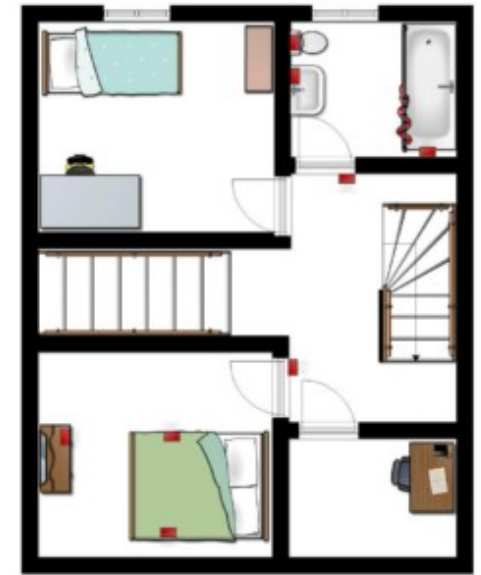
(a) House A



(b) House B



(c) House C, First floor



(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.

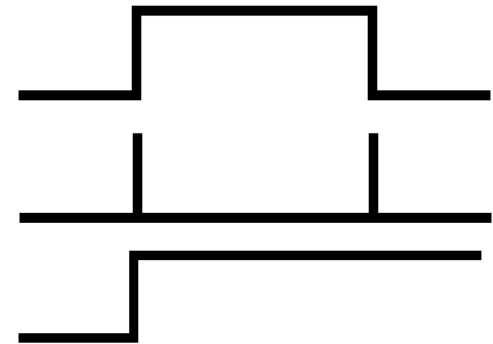
Methods

- 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

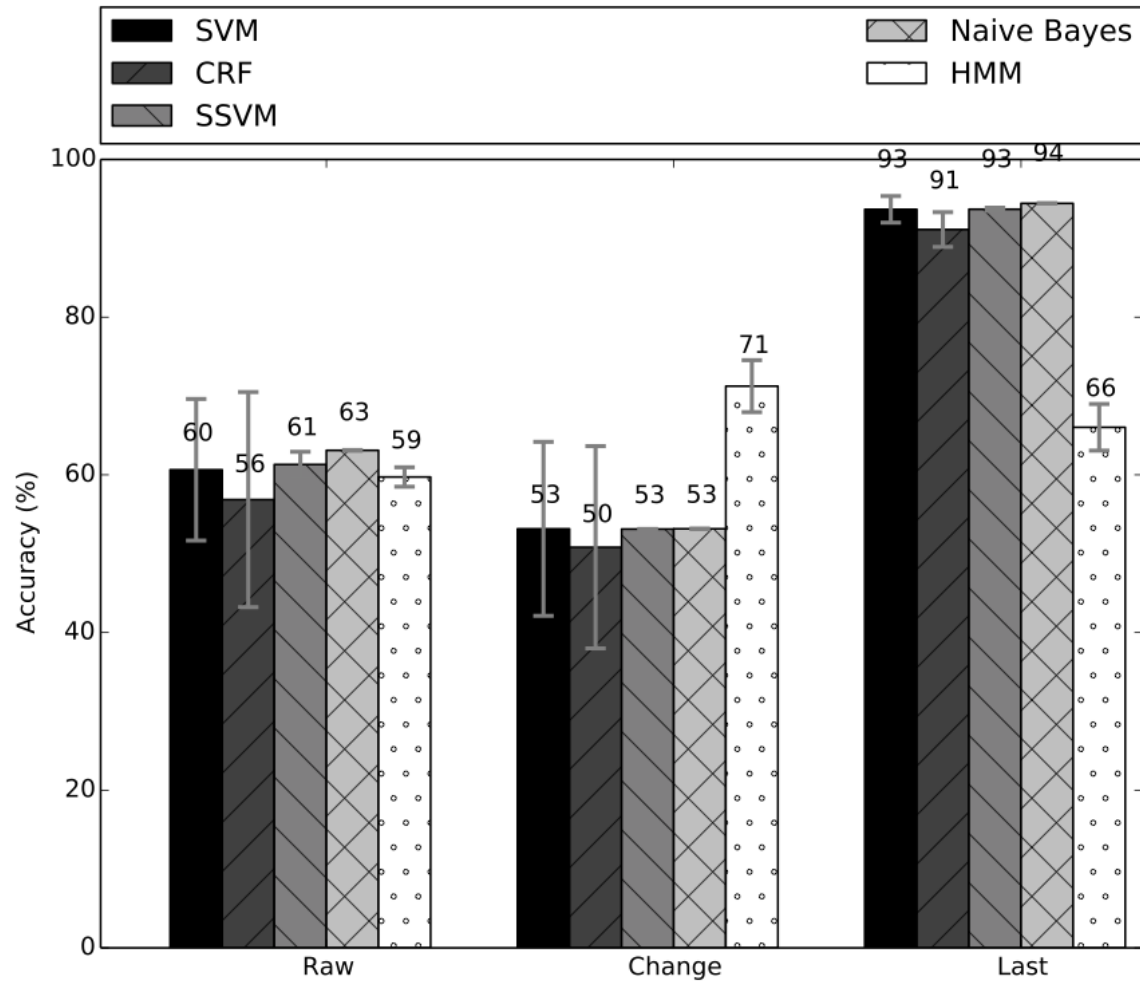
- Feature representations

- Raw
- Change
- Last



- 5 fold – cross validation
- Divide the dataset into smaller subsequences of ~2 hours

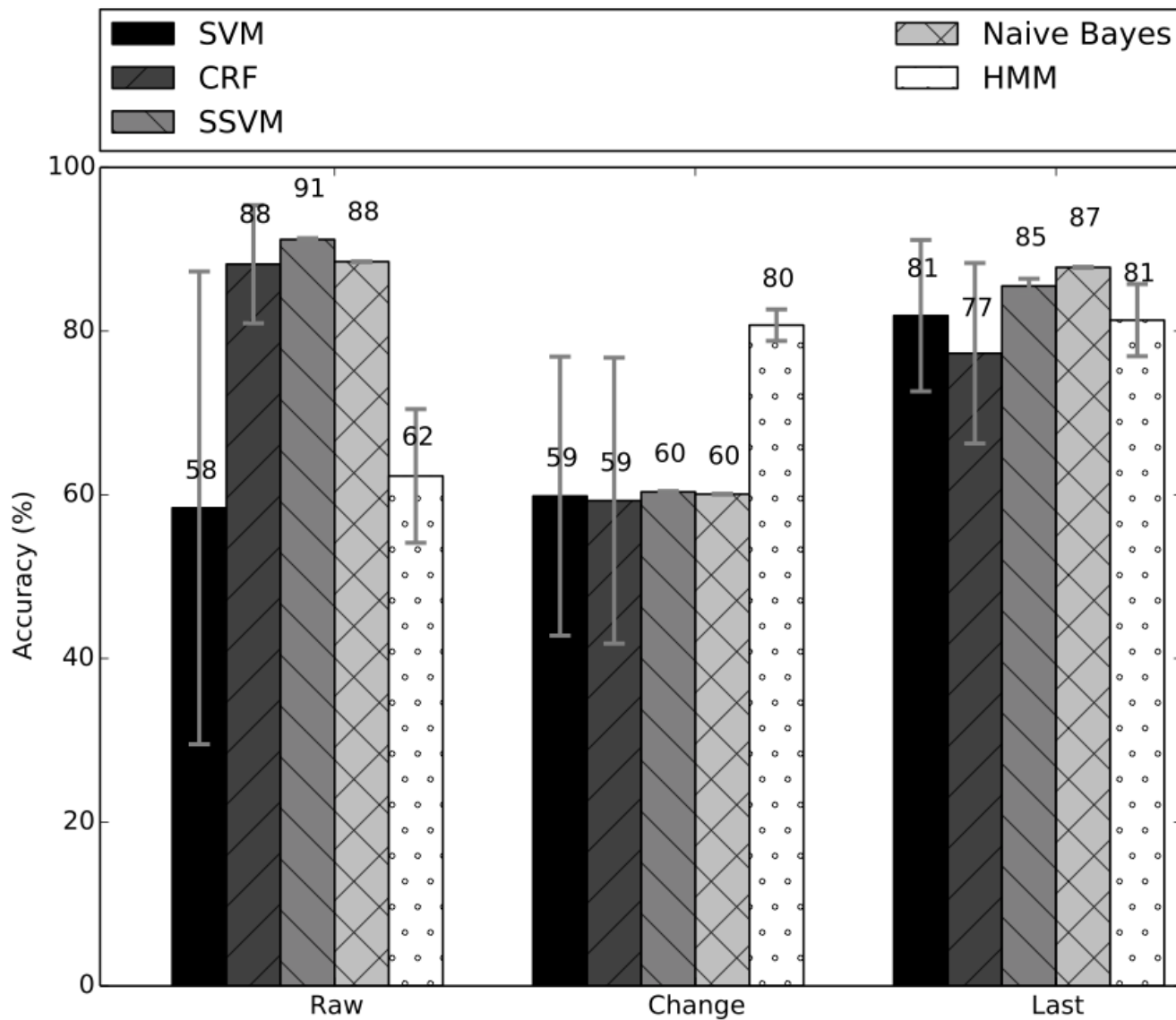
Comparison: House A



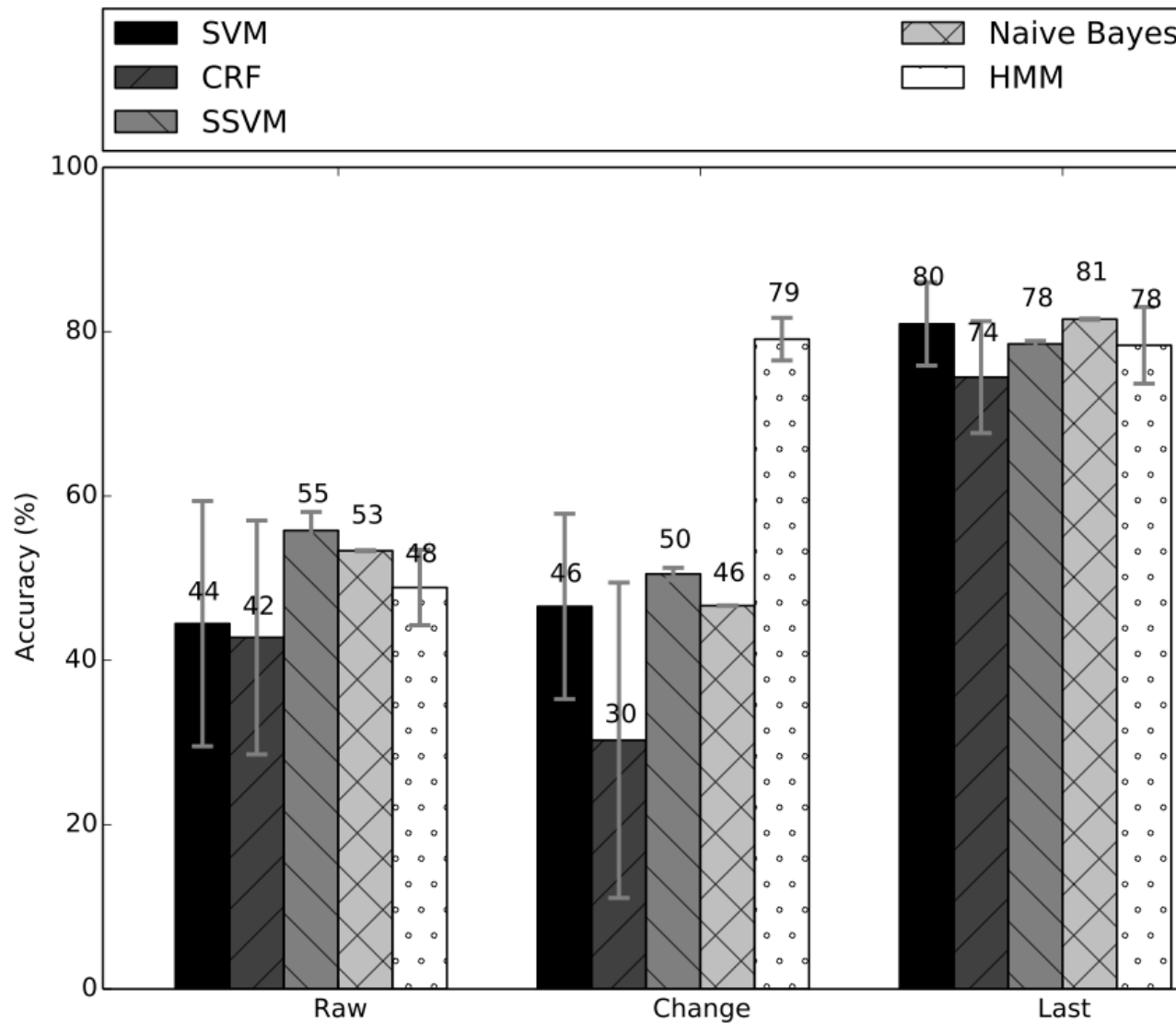
House A Metrics

| | | Precision | Recall | F-Measure | Accuracy |
|-------|---------|-----------|--------|-----------|-----------------|
| Model | Feature | | | | |
| 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 |
| 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.