

# Motivation

- Investigate the utility of mobile accelerometer data to identify human actions
- Learn data representations from mobile accelerometer using Restricted Boltzmann Machines (RBMs)
- Generate models to classify the learnt representations using Recurrent Neural Networks (RNN)

# Approach

• A set of unsupervised features are learnt, to recognize the phrases from American Sign Language (ASL), using RBM.



Fig 1: Overview: The stacked RBM network is trained for unsupervised feature generation. The computed features are used for phrase classification using RNN.

- Results are compared with the best performing supervised feature set [2].
- We created a dataset of 600 accelerometer readings collected from 50 users to validate the proposed approach.



# Talking Hands: RNN-based Sign Language Recognition Mohak Sukhwani (mohak.sukhwani@research.iiit.ac.in), Prateek Panwar (prateek.panwar@uoit.ca) **CVIT** International Institute of Information Technology ViaLab University of Ontario Institute of Technology

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| merry Christmas | We wish you a  | Predicted Phrase |
|-----------------|----------------|------------------|
| ;               | Y <sub>i</sub> |                  |



Fig 2: (Left) User wears the device on his wrist and performs the desired action. (Right) Figure shows the similarity in structure of accelerometer readings obtained while 4 different participants enact the given phrase. We intend to capture this similarity using our proposed solution.

# Applications

- Gestural phrase to speech conversion
- Translation to other languages
- Instant messaging
- Smart watch app



| Dataset        |  |                                  |  |  |  |  |  |  |  |
|----------------|--|----------------------------------|--|--|--|--|--|--|--|
| Name           | Contents   | Role                             |  |  |  |  |  |  |  |
| OPPORTUNITY[1] | Sampled wrist accelerometer data from the online dataset | RBM weights for feature learning |  |  |  |  |  |  |  |
| * ASL          | 600 accelerometer readings for selected phrases          | RNN weights for ASL recognition  |  |  |  |  |  |  |  |

\*The dataset is available online: http://bit.ly/talkinghands-data

| Results       |                |                |                |                |                |                |                |                |                |  |  |  |
|---------------|----------------|----------------|----------------|----------------|----------------|----------------|----------------|----------------|----------------|--|--|--|
| Method        | Raw Features   |                | BoW            |                | Method         | Raw Features   |                | BoW            |                |  |  |  |
|               | L <sub>1</sub> | L <sub>2</sub> | L <sub>1</sub> | L <sub>2</sub> |                | L <sub>1</sub> | L <sub>2</sub> | L <sub>1</sub> | L <sub>2</sub> |  |  |  |
| Naive<br>Base | 0.18           | 0.19           | 0.34           | 0.39           | SVM-<br>Linear | 0.13           | 0.13           | 0.16           | 0.20           |  |  |  |
| KNN           | 0.14           | 0.23           | 0.25           | 0.25           | SVM-Poly       | 0.28           | 0.25           | 0.325          | 0.30           |  |  |  |
| Reg. Trees    | 0.24           | 0.26           | 0.285          | 0.30           | SVM-RBF        | 0.19           | 0.19           | 0.45           | 0.46           |  |  |  |

Phrase classification accuracy using hand crafted features: L1 and L2 correspond to normalization scheme. The proposed unsupervised features obtain an accuracy of **0.53** with stacked RBMs which is **13%** more than the best performing supervised technique [2].



### References:

[1] Chavarriaga, Ricardo, et al. "The Opportunity challenge: A benchmark database for on-body sensor-based activity recognition." Pattern Recognition Letters 34.15 (2013): 2033-2042. [2] Zheng, Yonglei, et al. "Physical Activity Recognition from Accelerometer Data Using a Multi-Scale Ensemble Method." IAAI. 2013.



## Discussion

### Limitations:

-Present prototype uses only 1 hand -Limited phrase vocabulary

### Future work:

-Improved accuracy with more advanced movement capturing sensors -Expanded vocabulary -Bimanual Gestures