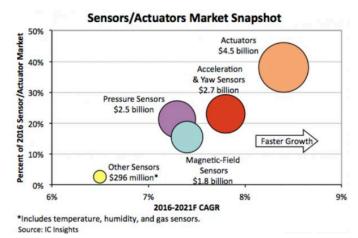
### **Conditional-UNet:** A Condition-aware Deep Model for Coherent Human Activity Recognition From Wearables

Liming Zhang, George Mason University Wenbin Zhang, University of Maryland, Baltimore County Nathalie Japkowicz, American University 12/10/2020

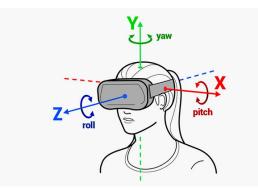


# Motivations

(1) Sensor-embedded wearables are more and more popular .etc.



https://www.eenewsanalog.com/news/sensor-actuator-markets-see-uptick-wearables-embedded-control-and-iot/page/0/1



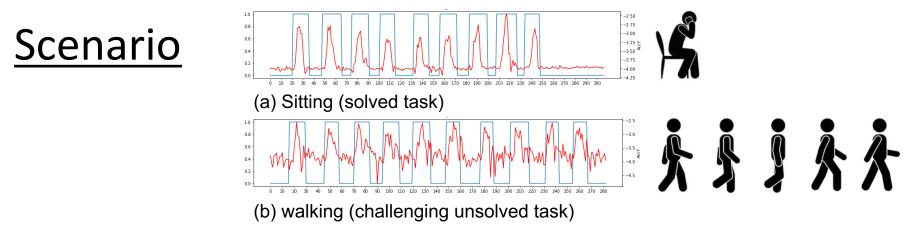
(2) Head gesture recognition with wearables are a trending research.

https://www.veative.com/blog/gyroscope-importantvirtual-reality/

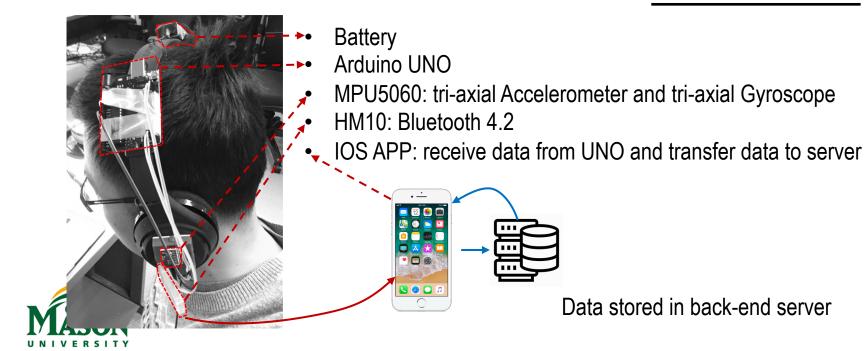
(3) Real world conditions can be complicated when users are moving.
None of current works tried to solve this.



Wu, Cheng-Wei, et al. "Applying machine learning to head gesture recognition using wearables." 2017 IEEE 8th International Conference on Awareness Science and Technology (iCAST). IEEE, 2017.



## Hardwares



Data stored in back-end server

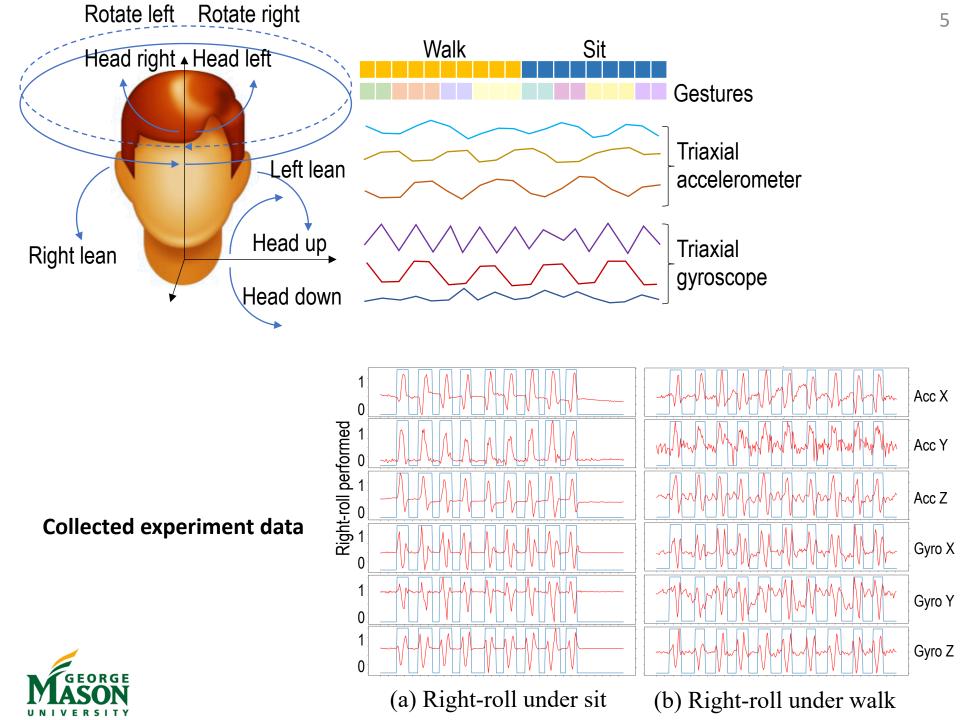
## Challenges

A new problem: *Coherent* Human Activity Recognition(Co-HAR) with *single-location* sensors

Specifically, there are modeling challenges as follows:

- 1. The single location of sensors has mutual impact on signals.
- 2. The imbalanced domination of different activities could fade away the signals of the other activities.
- 3. The multi-label window problem for activities of various duration





## A novel condition-aware deep model called "Conditional-UNet"

Raw data likelihood formula

 $p(Y_1, \dots, Y_H | X) =$   $p_{\theta_1}(Y_1 | X) p_{\theta_2}(Y_2 | Y_1, X) \dots p_{\theta_H}(Y_H | P_{H-1}, \dots, Y_1, X)$ (1) Yi: different labels X: sensors data

**Our approach:** Conditional data likelihood factorization as a more general framework

$$\mathcal{L} = log(p(Y_1, \dots, Y_H | X)) = \sum_{t}^{T} \left( log(p_{\theta_1}(Y_{1,t} | X)) + \dots + log(p_{\theta_H}(Y_{H,t} | P_{H-1,t}, \dots, Y_{1,t}, X)) \right)$$
(2)

**Existing approaches:** multi-label classification assuming conditional independences

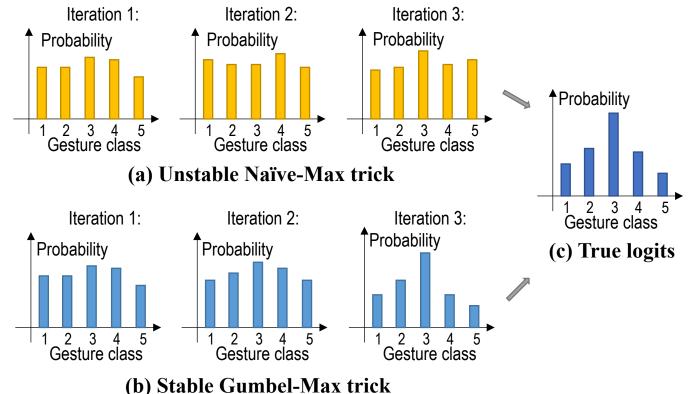
$$p(Y_1, \dots, Y_H | X) = p_{\theta_1}(Y_1 | X) p_{\theta_2}(Y_2 | X) \dots p_{\theta_H}(Y_H | X)$$

## A novel condition-aware deep model called "Conditional-UNet"

#### Unet used a 32 hidden neuron for Deep architectures the first upsampling and last downsampling layers, and is very Decoding module: $f_{\theta_1}(X)$ efficient for computing. $\hat{y}_1$ UNet Encoding module: $g_1(\hat{y}_1)$ $Merge(X,\cdot)$ $\widehat{Y}_1$ $\Theta = minimize(\mathcal{L})$ Embedding 4 Optimizing module X $Embed(\hat{Y}_1)$ **Generate** $(\hat{y}_1)$ Decoding module: $f_{\theta_2}(Y_1, X)$ Encoding module: $g_2(\cdot)$ 4.77 $\hat{Y}_2$

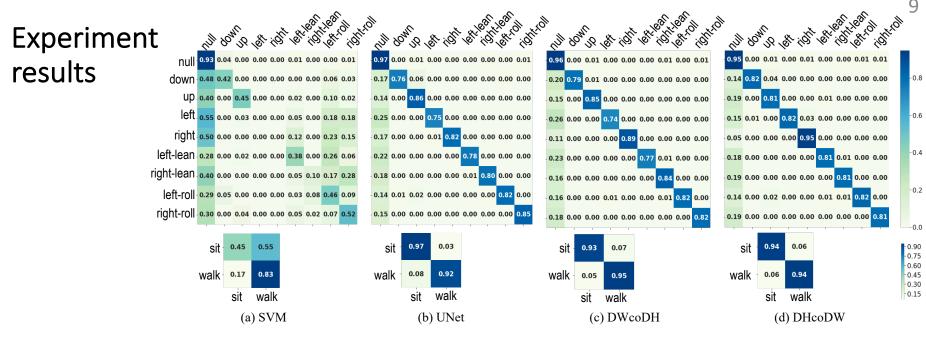
## A novel condition-aware deep model called "Conditional-UNet"

Compared Naïve-Max and Gumbel-Max trick for **Generate** $(\hat{y}_1)$   $\mathcal{M}$ sampling operation





(b) Stable Gumbel-Max trick



True Head Label

right-roll left-roll right-lean

left-lean

right left

up down

nul walk

sit

5 10 15

(a) SVM

Predict Head Labe

10 15

(b) UNet

0 5

10 15

0 5 10

(c) DWcoDH (d) DHcoDW

0 5

True Walk/Sit

Predict Walk/Sit

Acc X

Acc X

Baseline models: SVM, UNet

Two alternative models of Conditional-Unet:

- 1) DWcoDH, Walking conditioned on Head
- 2) DHcoDW, Head conditioned on Walking

Conditional-UNet outperforms existing state-of-the-art UNet model, and achieves up to **92.06%** of accuracy and **87.83%** of F1 score. Also, DHcoDW's performance is good for all gesture types, not for some gestures like other baselines.



# Contributions, Limitations and Future works

Contributions:

Addressed a challenging problem Co-HAR for which a new dataset was collected.

Proposed a novel condition-aware deep model called "Conditional-UNet".

#### Limitations:

- It is still not real-life scenario.
- Need to include more deep learning methods to compare.
- We run deep models on desktop GPU, but computation power is constrained in real-world wearables.

#### **Future works**

- Is such trained model transferred for real-world scenario? Or need re-training?
- In the data likelihood loss, hierarchical labels can be considered or imbalanced class problem can be studied in the future.



**Code and data:** https://github.com/tongjiyiming/Conditional-UNet