

# Smart Wearable Controlling System by Hand and Finger Gestures

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# Content

- Introduction
- System Design
- Method
- Experiments and Results
- Conclusion & Future Work

# Content

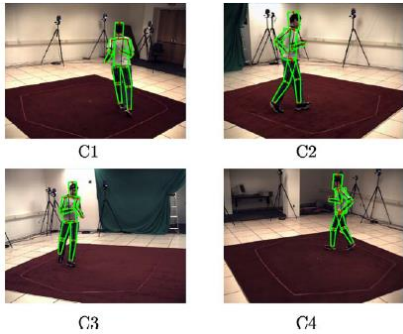
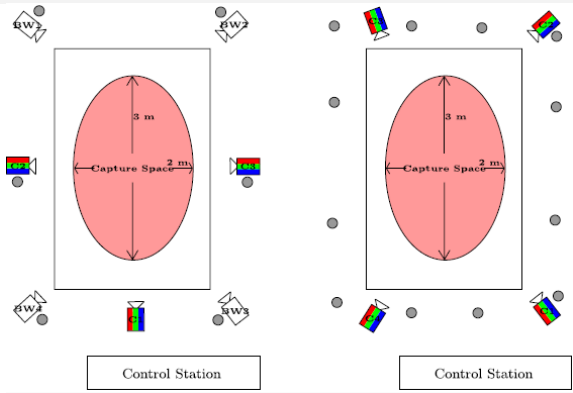
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# Gesture Recognition

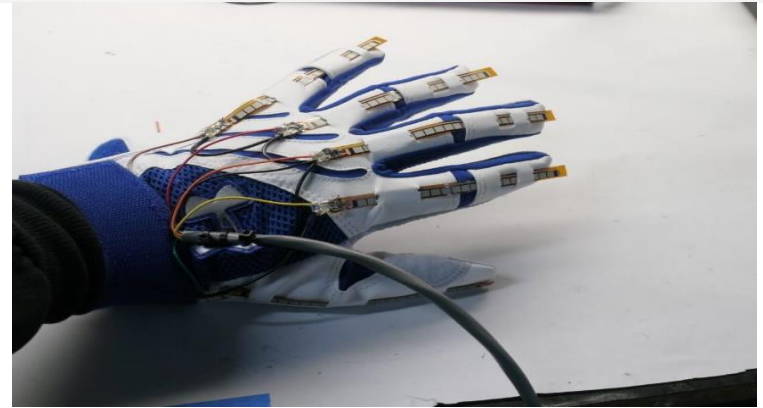
- Gesture recognition has become one of the popular research fields that served as an intelligent and a natural interface.
- Exploring alternative ways to recognize the imitate human gestures to perform complex and difficult tasks in various areas.
- Wearable computing trends allowed devices to be portable and smaller in size.



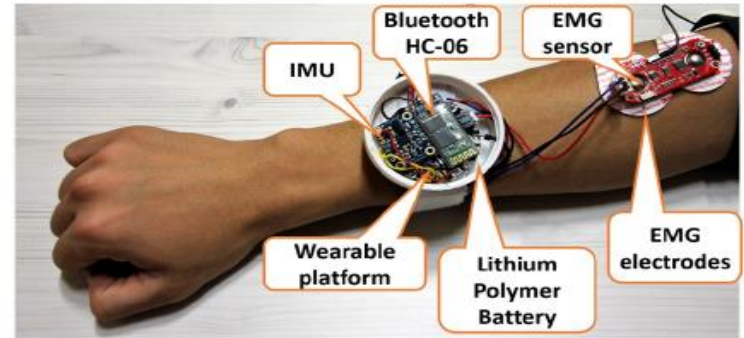
# Literature Review



Vision system [1] captures the motions by using multiple high-speed infrared cameras



Pathk [2] presented a framework for hand gesture recognition



Gesture control armband using single EMG and IMU

# Smart Wearable Controlling System

Tend to wearable computing trend and high accuracy rate:

- Utilizing only three IMUs
- Recognizing 22 gestures



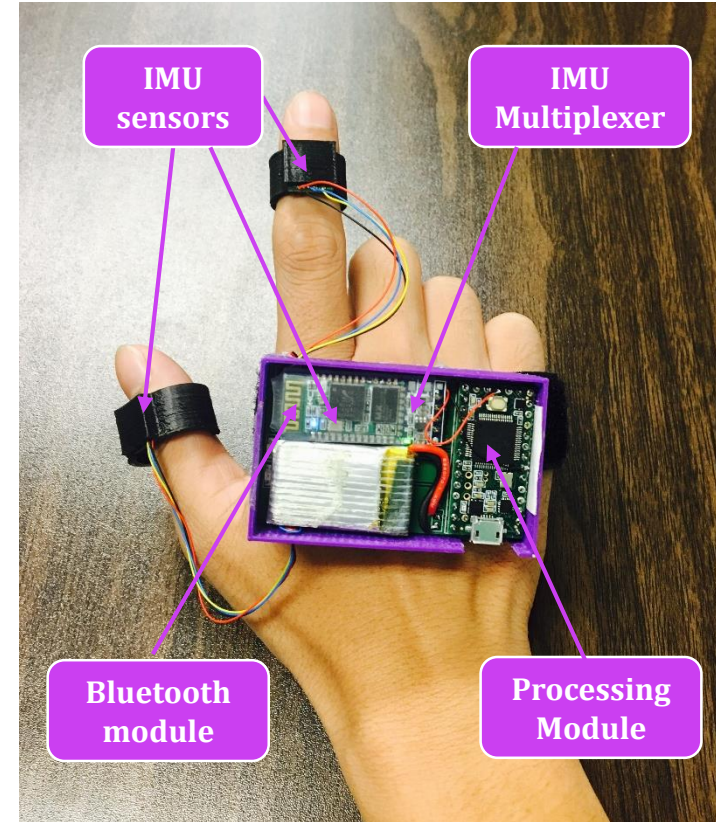
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# System Overview

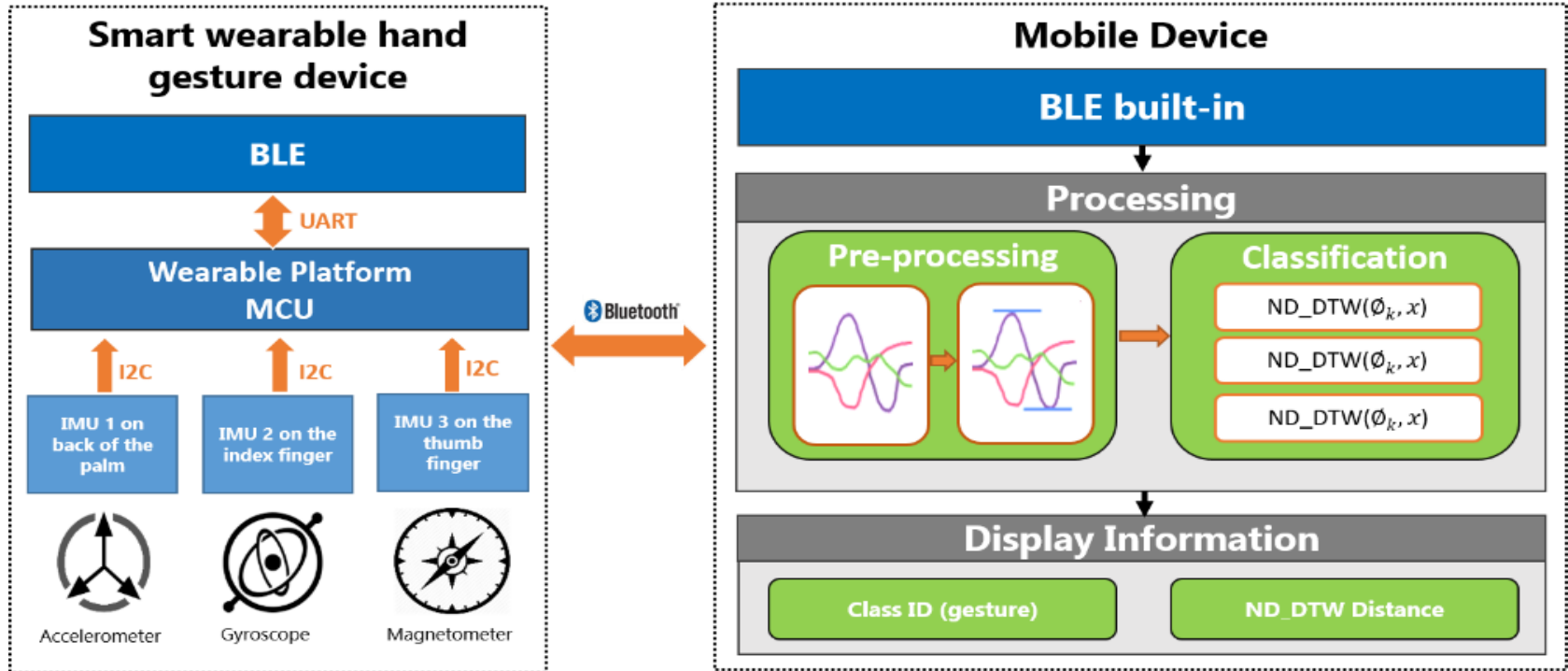
The system includes:

- Sensor module
- Processing module
- Mobile device





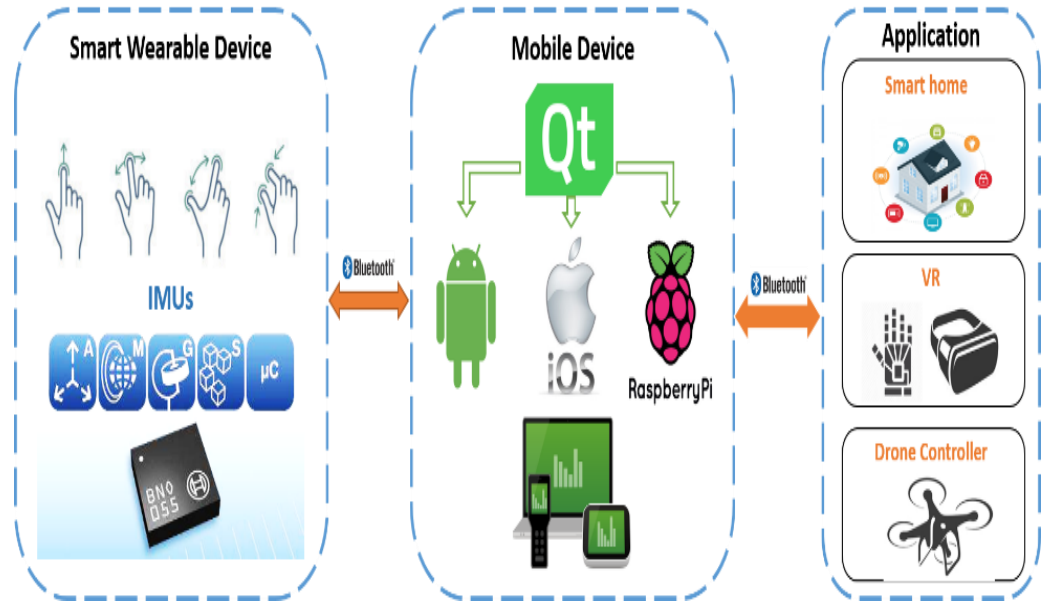
# System Overview



System design overview.

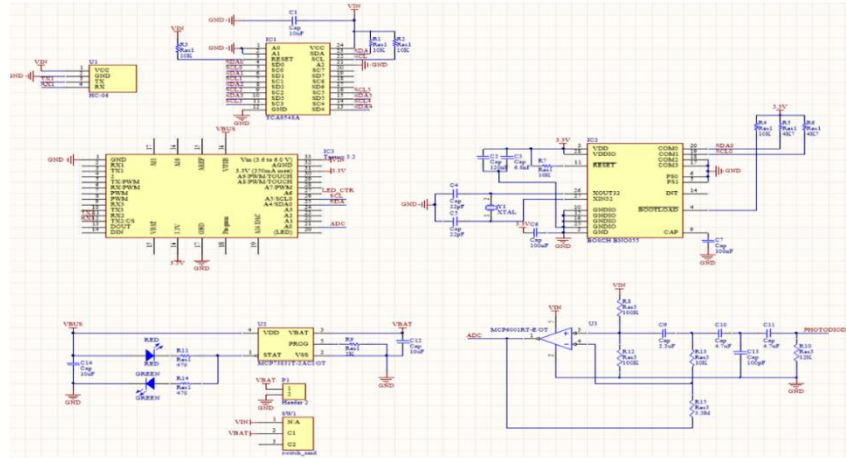
# System Overview

- Qt cross platform software
- Applying to various applications as home automation, games controller ...

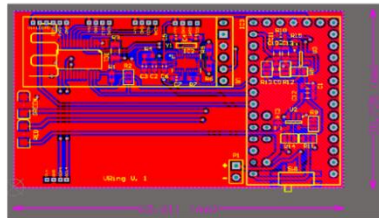


Proposed smart wearable controller system can be applied to various applications.

# Hardware

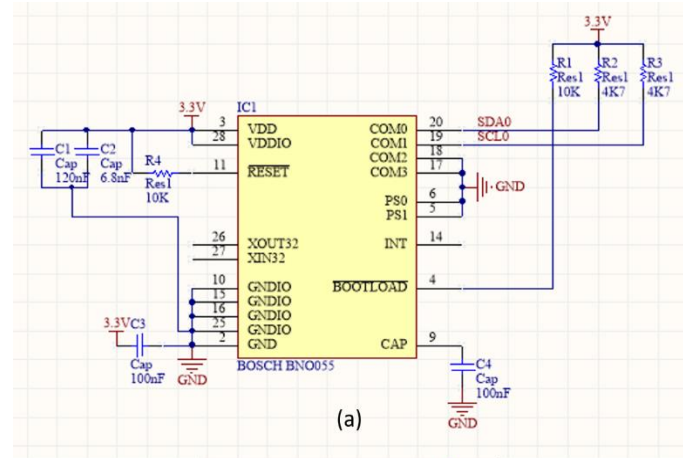


(a)

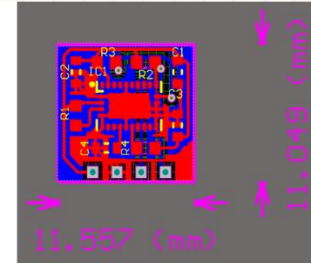


(b)

(a) Schematic and (b) PCB layer design of processing module.



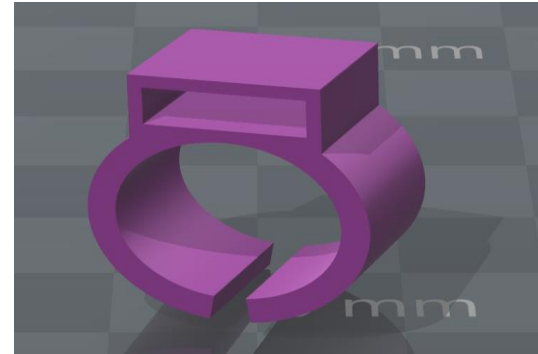
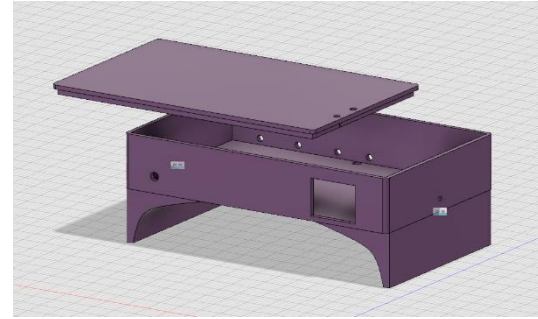
(a)



(b)

(a) Schematic and (b) PCB layer design of sensor module.

# Hardware



Appearance design using 3D Printer

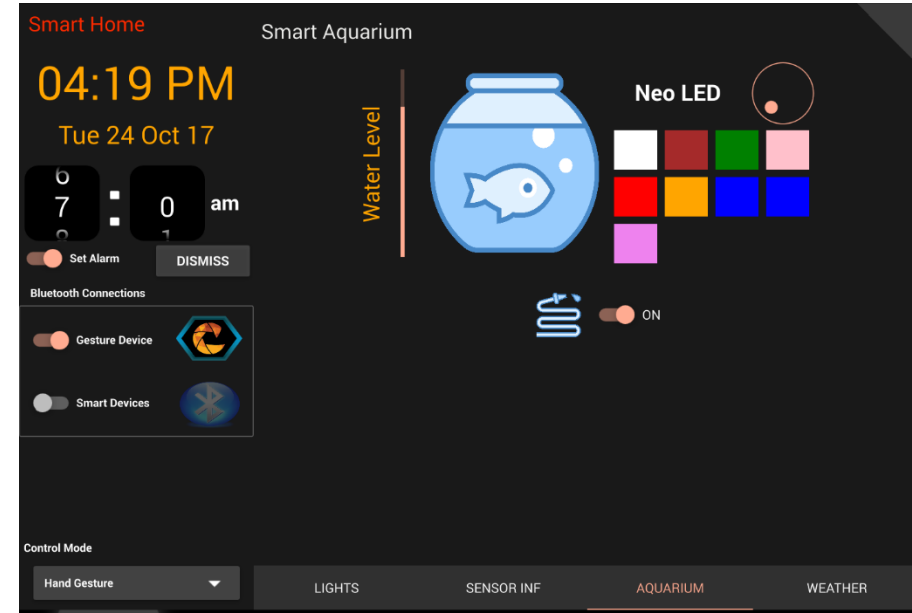
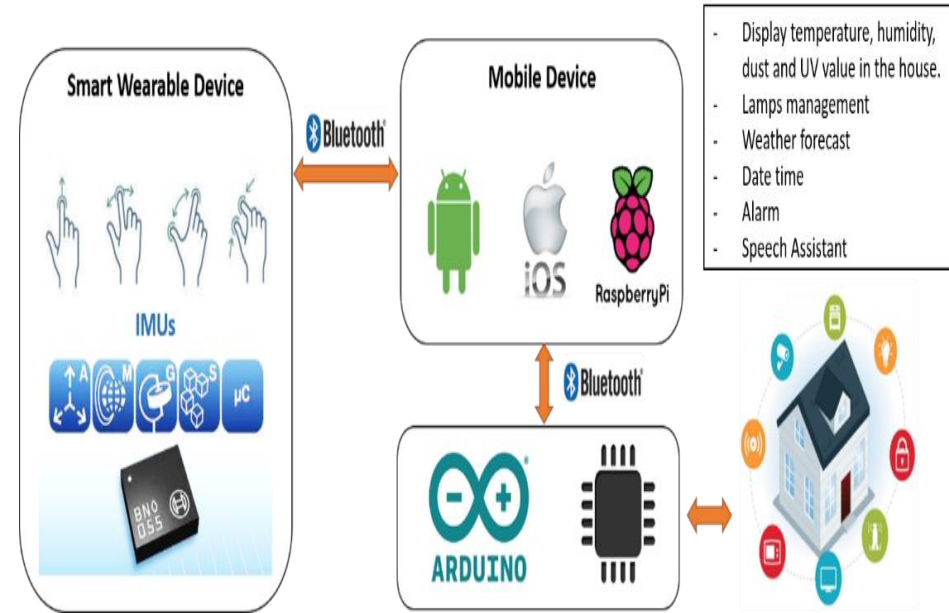
# Design of Software

- Cross-platform application framework (Linux, OS X, Windows, Android, iOS).
- Framework written in C++.
- GUI is written in QML instead of the whole application whereas the application's backend is written in C++.



Qt cross-platform application development framework.

# Design of Software



System overview of smart home application.

Screenshot of the smart home application

# Design of Software

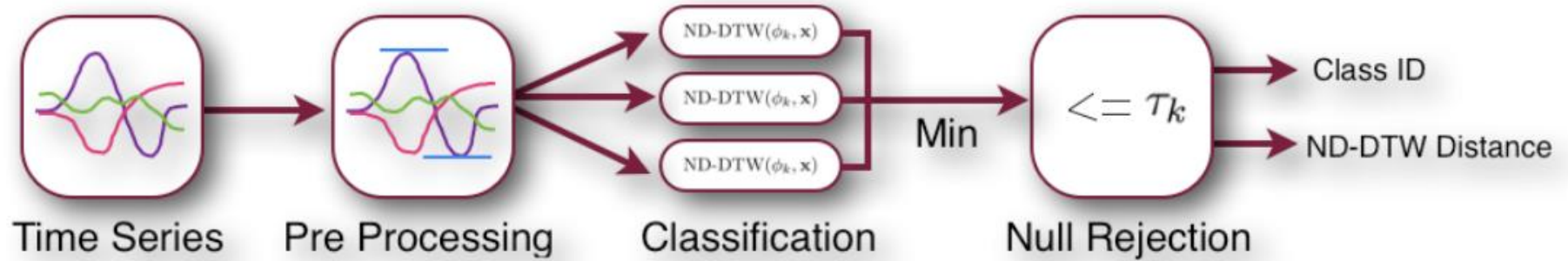
Video Demo

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  - N Dimension – DTW
  - Pre-processing
  - Training
  - Classification Threshold
- Experiments and Results
- Conclusion & future work



# Method

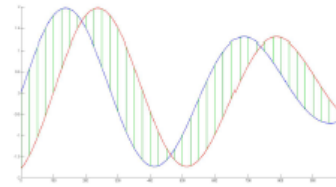


Flowchart of method

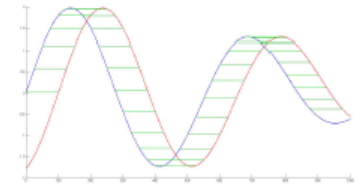
# Dynamic Time Warping (DTW)

DTW is an algorithm that can compute the similarity between two time-series, even if the lengths of the time-series do not match.

DTW has been applied to temporal sequences of video, audio, graphics data and gesture recognition.



(a) Euclidean Distance



(b) Dynamic Time Warping

Two identical time-series signals, with phase differences measured with (a) Euclidean distance and (b) Dynamic Time Warping.

# Dynamic Time Warping (DTW)

- Given two, one-dimensional, time-series:

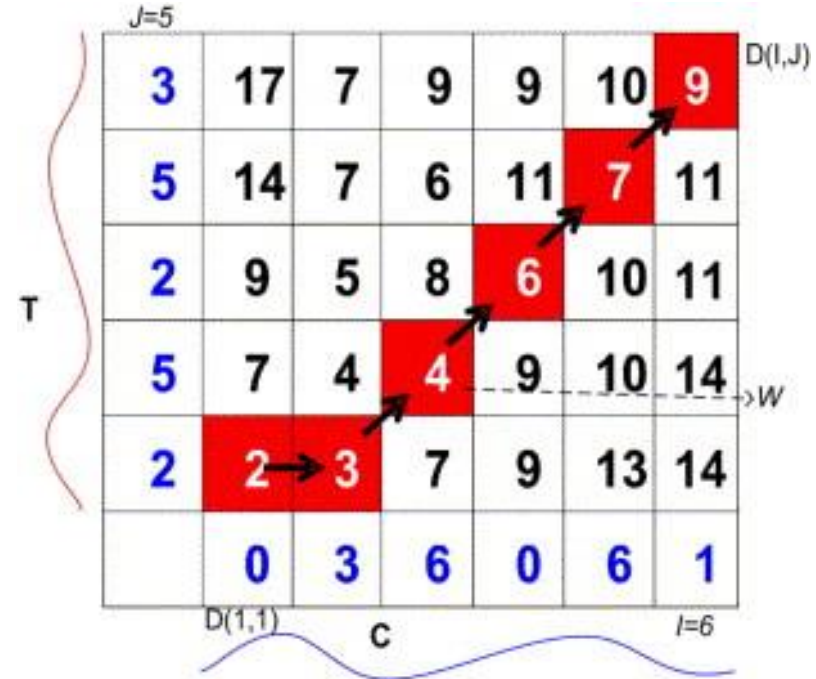
$$x = \{x_1, x_2, \dots, x_{|x|}\}^T$$

$$y = \{y_1, y_2, \dots, y_{|y|}\}^T$$

- Construct a warping path  $w = \{w_1, w_2, \dots, w_{|w|}\}^T$
- A number of constraints are placed on the warping path which are as follows:
  - start at:  $w_1 = (1, 1)$
  - end at:  $w_{|w|} = (|x|, |y|)$
  - Continuous
  - cannot move backwards.
- The warping cost given by:

$$\min \frac{1}{|w|} \sum_{k=1}^{|w|} DIST(w_{k_i}, w_{k_j})$$

where  $DIST(w_{k_i}, w_{k_j})$  is the distance function (Euclidean)

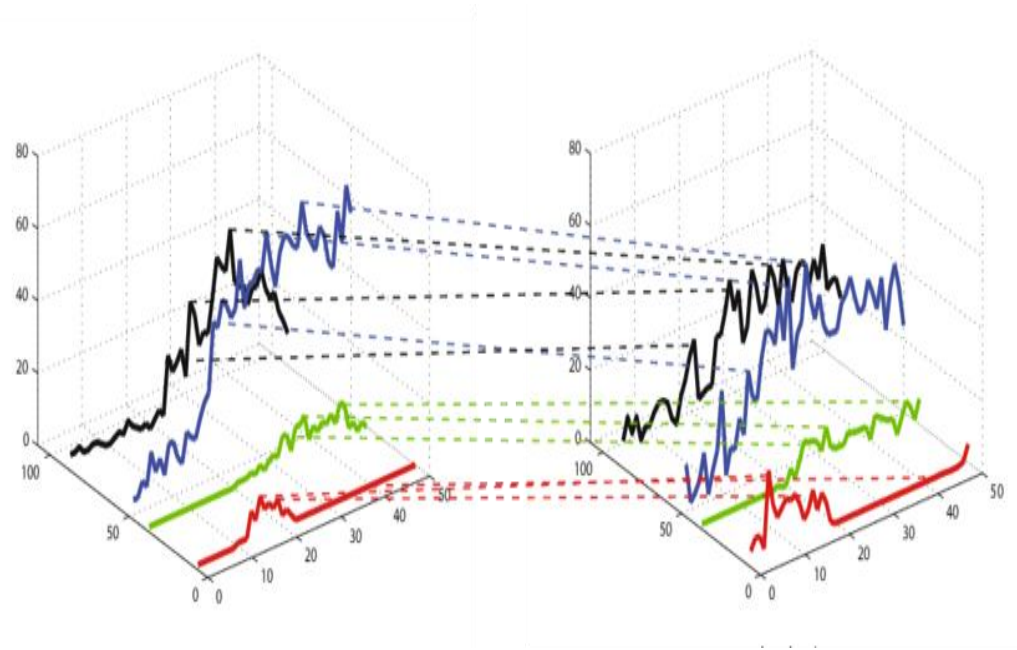


Example of warping path in a distance table. Length of class sequence,  $x = 6$  and length of test sequence,  $y = 5$ .

# N Dimension DTW

- Computing the distance between 2 given time-series.
- The total distance across all N dimensions is used to construct the warping matrix with the Euclidean distance:

$$DIST(i,j) = \sqrt{\sum_{n=1}^N (i_n - j_n)^2}$$



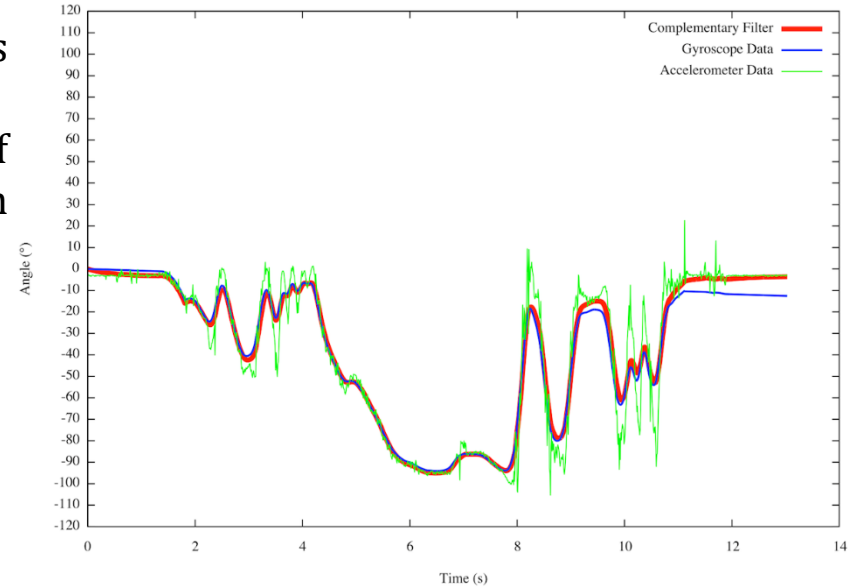
Multiple dimension DTW

# Pre-processing

- Accelerometer is reliable only in the long run, but so noise in the short run.
- Gyroscope is reliable only in the short run, but is drifted in the long run
- Using measurement by the fusion of accelerometer and gyroscope data with Complementary filter:

$$\text{angle} = (0.98) * (\underbrace{\text{angle} + \text{gyro} * \text{dt}}_{\text{Integration.}}) + (\underbrace{0.02 * (\text{x\_acc})}_{\text{Low-pass portion acting on the accelerometer.}});$$

Something resembling a high-pass filter on the integrated gyro angle estimate. It will have approximately the same time constant as the low-pass filter.



# Training

- Template is required to store all the intended gestures in database by recording  $M_g$  training samples for each of the  $G$  gestures
- Each of  $G$  template can be found by computing distance each of  $M_g$  training samples
- Searching for the training samples that provide the minimum warping distances when compare the other training samples in that class

# Classification Threshold

- The N-dimensional input time-series X is not matched with any of the gestures in the template
- Thus this false classification issue can be resolved by defining a classification threshold for each gesture in the template during the training phase.
- The algorithm will indicate the particular gesture as no match:

$$c = \begin{cases} c & \text{if } (d \leq \tau_g) \\ 0 & \text{otherwise} \end{cases}$$

where d is the total normalized warping distance

$$\tau_g = \mu_g + (\sigma_g \gamma)$$
$$\mu_g = \frac{1}{M_g - 1} \sum_{i=1}^{M_g} 1 \{ND - DWT(\phi_g, X_i)\} \quad \sigma_g = \sqrt{\frac{1}{M_g - 1} \sum_{i=1}^{M_g} 1 \{ (ND - DWT(\phi_g, X_i) - \mu_g)^2 \}}$$

$\gamma$  could be initially set to any number during the training phase and later be adjusted by the user

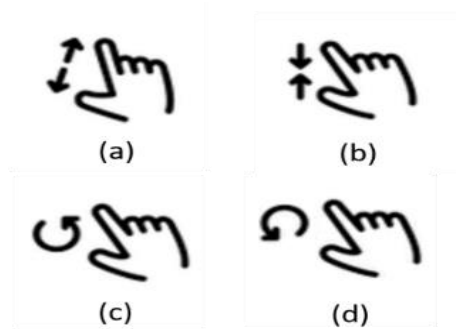
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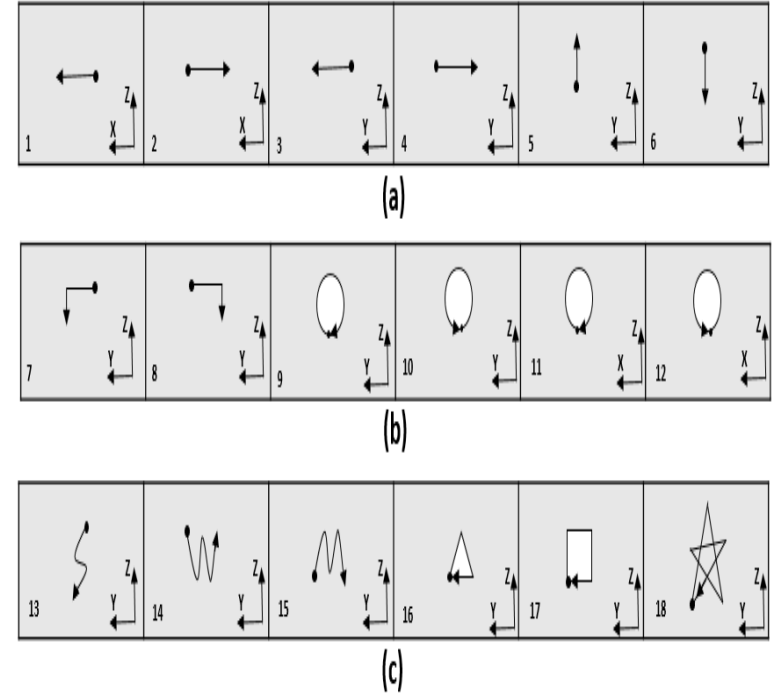


# Experiment Setup

- Six right handed subjects are voluntarily participated in the experiments.
- Subjects are requested to perform 25 repetitions of 22 gestures.
- The gestures consisted of 18 hand gesture and 4 fingers gestures.

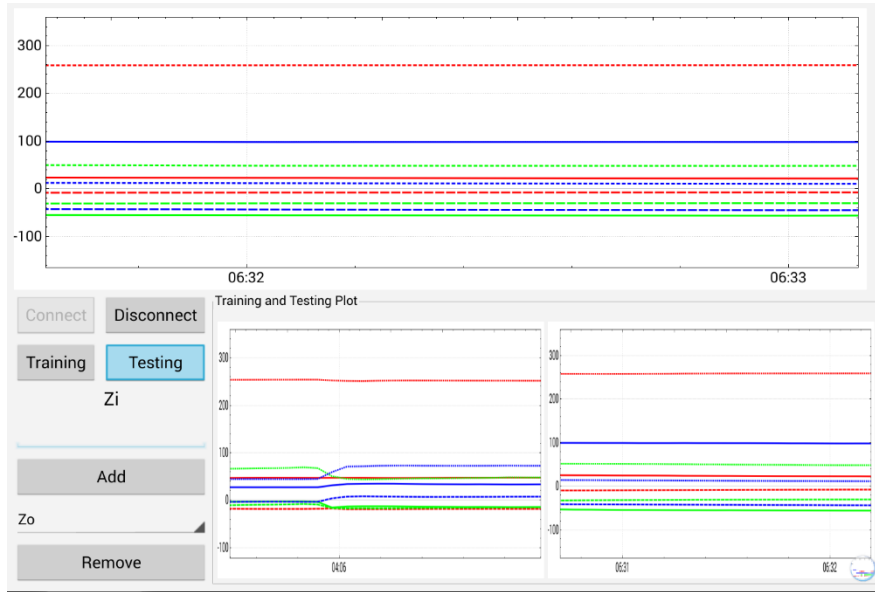


Four finger gestures

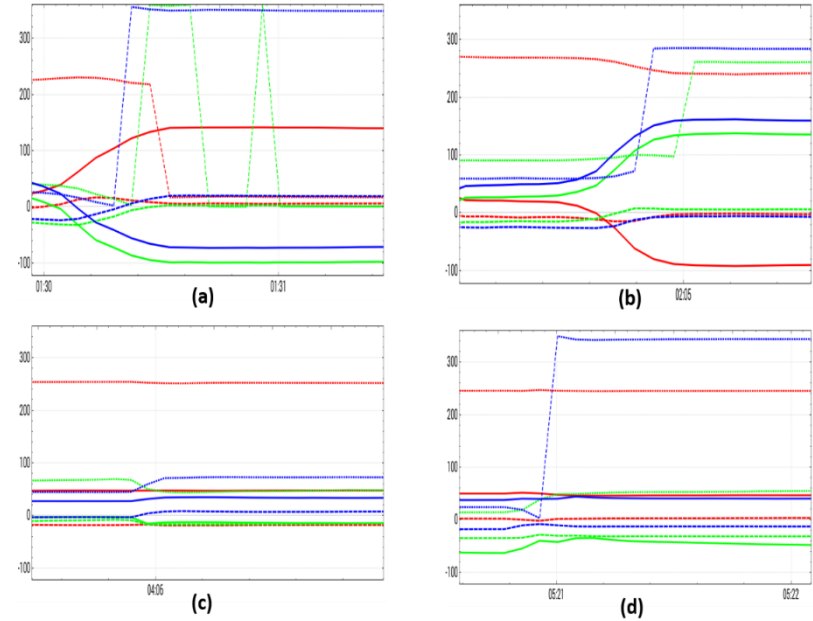


The predefined 18 gestures

# Experiment Setup



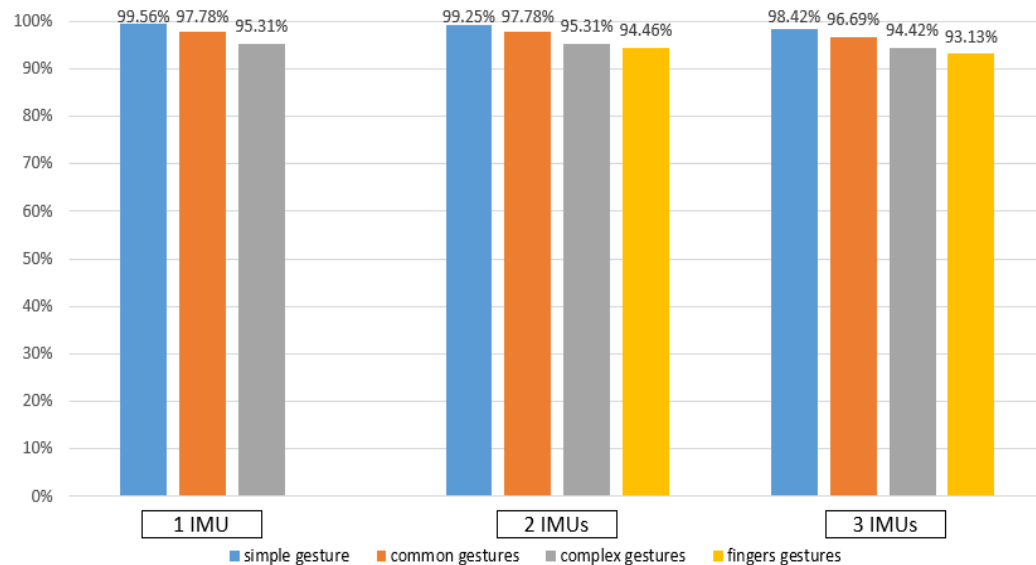
Self-development application for the purposes of training phase and testing phase.



Orientation samples obtained from 3 IMUs to the gestures including (a) rotate right, (b) rotate left, (c) zoom in and (d) zoom out.

# Result and Discussion

- Experiments are conducted to validate the gesture recognition accuracy rate utilizing the ND-DTW algorithm.
- The impact to accuracy rate based on the quantity of IMUs used are also taken into consideration.
- The ND-DTW model was trained using 6-fold cross validation method



# Result and Discussion

Table 4-4 Comparison of related systems.

Author	Type of sensors	Quantity of sensors	Number of gesture	Accuracy rate	Paper
Chang	IMU EMG	3	6	95.6%	[39]
<u>Georgi</u>	IMU EMG	17	12	97.8%	[40]
<u>Khezri</u>	EMG	4	6	92%	[41]
<u>Englehart</u>	EMG	4	6	98%	[42]
Liu	IMU	1	8	98.6%	[13]
Proposed study	IMU	1	18	97.55%	
Proposed study	IMU	2	22	96.90%	
Proposed study	IMU	3	22	95.89%	

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  - Conclusion
  - Future Work

# Conclusion

- Focused on the design and implementation of smart wearable controller system using hand and fingers for gesture recognition.
- The system utilized 3 IMUs sensors, to recognize total of 22 pre-defined gestures with DTW algorithm with low cost implementation and high accuracy rate.
- The study evaluated the IMUs quality impacted to the accuracy rate of system.
- System performed well with low energy consumption of 10 hours with 3.7V 240mA battery

# Future Work

- Optimization of PCB design in term of size and dimensionality to reduce power consumption and its wearable comfortability.
- Implementation of recognition algorithm in embedded hardware as standalone system which is compatible to different software platforms.
- Recognition of more complex gestures that involved three dimensional gestures by utilizing hand and fingers motion.
- Inclusion of more functionalities for various applications and allow gestures self-customization by users.

# THANK YOU!



# Q&A

## Reference