

Sign Talk

A Sign Language Recognition System

Sign Talk

- sEMG vs. RGB-D
- 10 English alphabets with distinct signs were chosen

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Background Work for SLR

Gloves

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- Users wear a glove which can detect movement and acceleration
- Allows the ability to easily work with a wide variety of hand sizes
- Complex machine learning techniques were not required to handle recognition from gloves
- Able to get accuracy of ~99% for letter by letter recognition
- Unable to do recognize complex gestures requiring significant movement

Camera + Depth Sensor

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Camera

- Without depth sensors and only 1 camera letter by letter recognition as done by reasonable success
- Convolutional models were used
- Transfer learning from other big image recognition frameworks was also effective

Camera + Depth

- Most published research used microsoft's kinect for depth sensor
- Was able to get 99% on letter recognition
- With multiple cameras, word by word recognition was successfully achieved with a dictionary of a few hundred words

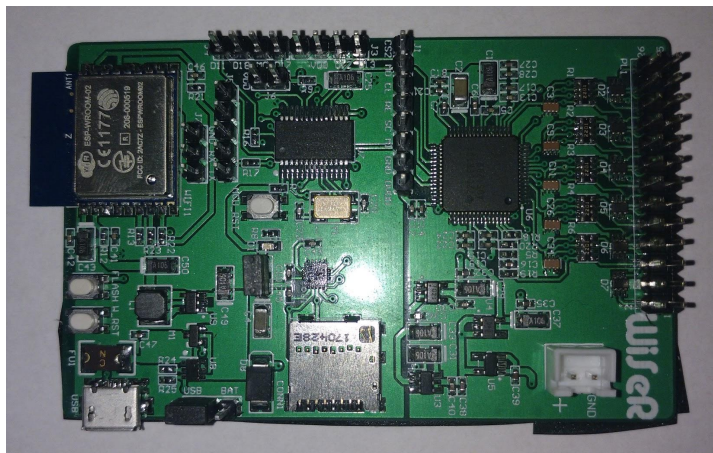
Wearables such as EMG sensors

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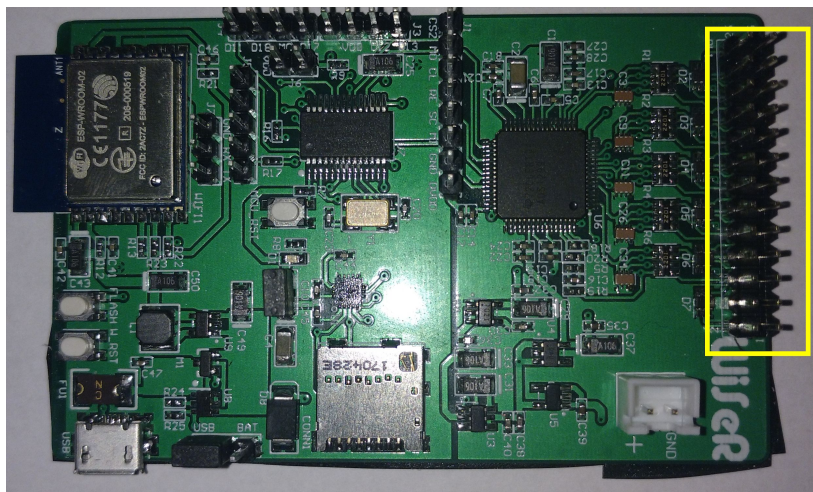
- Devices include myo or custom board with emg sensors
- Published research was able to successfully recognize character by character translation
- Difficult to do word by word translation due to the interactions of other muscles in the arm
- algorithms such as SVM produced reasonably well for this task

Hardware

- ADS1299
- PIC32MX350F128D
- ESP-WROOM-02

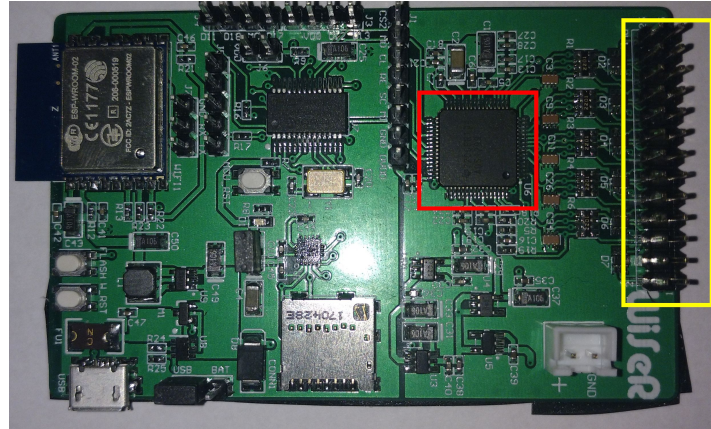


EMG - Channels



| PL1 | | | | | |
|---------|----|----|----|----|------|
| AVDD | 1 | 1 | 2 | 2 | AVDD |
| GND | 3 | 3 | 4 | 4 | GND |
| BIAS | 5 | 5 | 6 | 6 | BIAS |
| 8N | 7 | 7 | 8 | 8 | 8P |
| 7N | 9 | 9 | 10 | 10 | 7P |
| 6N | 11 | 11 | 12 | 12 | 6P |
| 5N | 13 | 13 | 14 | 14 | 5P |
| 4N | 15 | 15 | 16 | 16 | 4P |
| 3N | 17 | 17 | 18 | 18 | 3P |
| 2N | 19 | 19 | 20 | 20 | 2P |
| 1N | 21 | 21 | 22 | 22 | 1P |
| SRB2 | 23 | 23 | 24 | 24 | SRB1 |
| AVSS | 25 | 25 | 26 | 26 | AVSS |
| CON2X13 | | | | | |

EMG - PCB

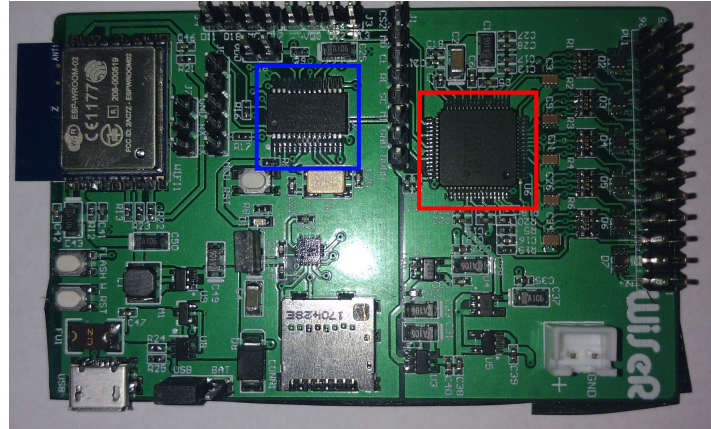


Yellow: Channels

Red: 24-bit Analog-to-Digital Converter

EMG - PCB

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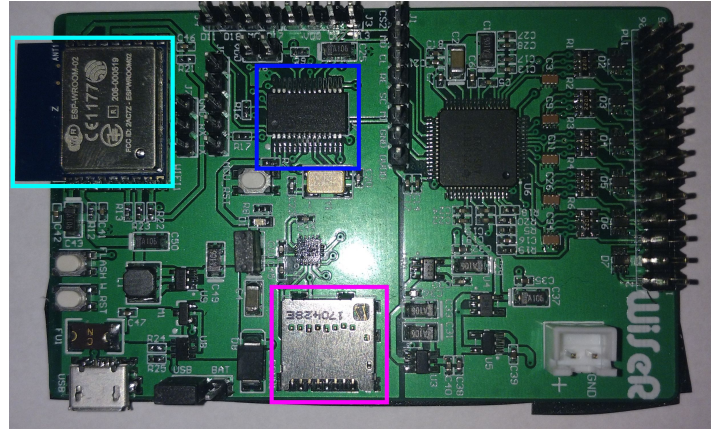


Red: 24-bit Analog-to-Digital Converter

Blue: Microchip

EMG - PCB

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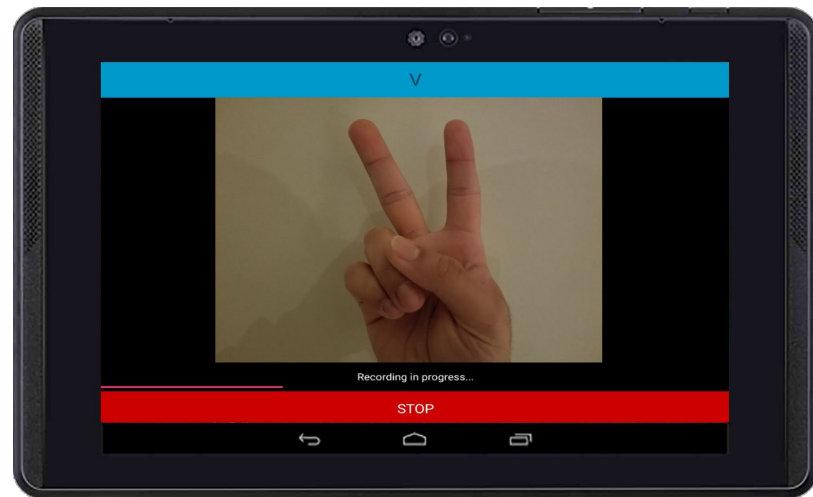


Blue: Microchip

Magenta: Secure Digital Card mount

Cyan: Wireless network card

Tango Tablet



Implementation

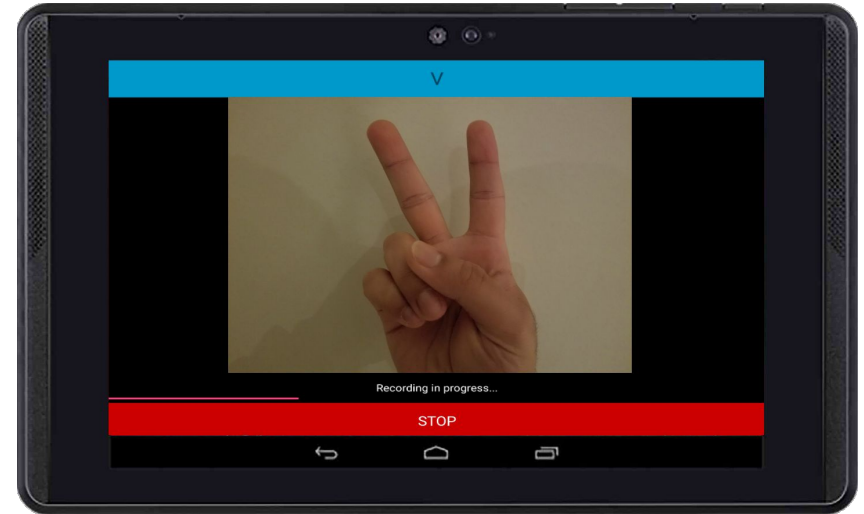
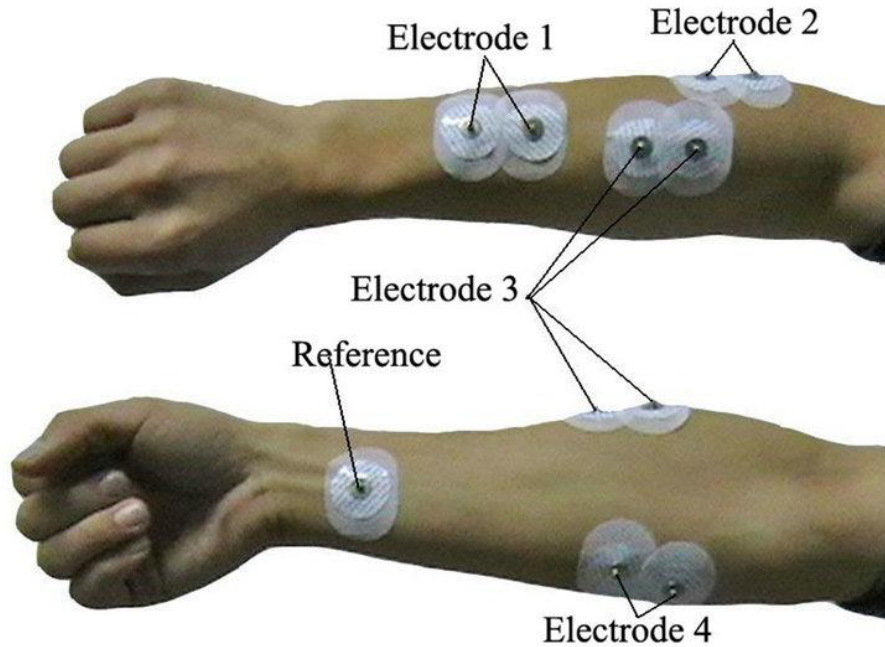
Process

1. Data Collection
2. Preprocessing
3. Feature Extraction
4. Classification

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Data Collection

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Preprocessing

Zero Mean

$$X_{\text{zero}} = X - \text{mean}(X)$$



<https://upload.wikimedia.org/wikipedia/commons/1/1d/Medianfilterp.png>

Feature Extraction (sEMG)

$$MAV = \frac{1}{N} \sum_{i=1}^N |Xi|$$

$$SSI = \sum_{i=1}^N |Xi|^2$$

$$RMS = \sqrt{\frac{1}{N} \sum_{i=1}^N Xi^2}$$

$$STD = std(X)$$

$$MAX = max(X)$$

$$MIN = min(X)$$

Feature Extraction (RGB-D)

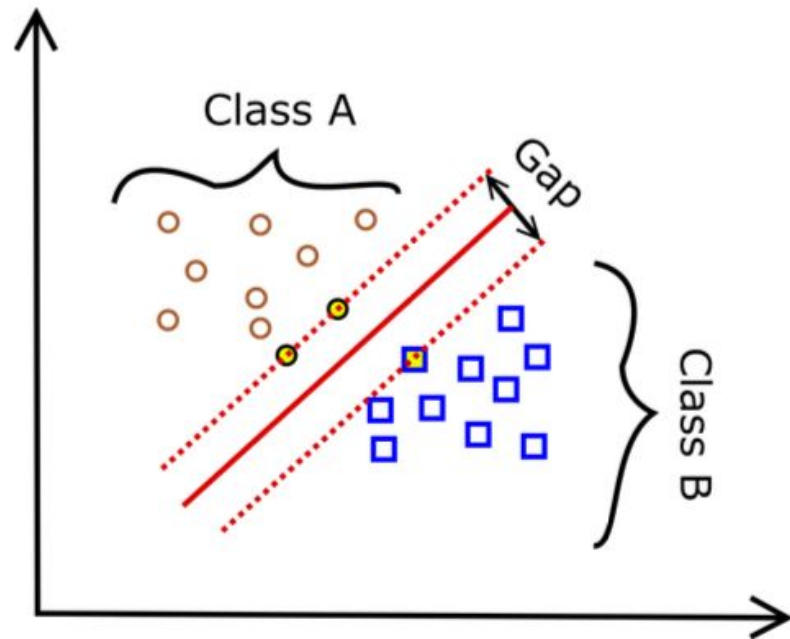
- Histogram of oriented gradients



<https://www.learnopencv.com/wp-content/uploads/2016/12/histogram-of-oriented-gradients.jpg>

Classification

- Support Vector Machine (SVM)
- Independent Component Analysis (ICA)
- 3D Convolutional Neural Network



Results

Unknown

- Issues with data collection
- Limited documentation on Hardware
- Couldn't find relevant datasets

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References

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- [1] Ferat Sahin Celal Savur. “Real-time American Sign Language Recognitionwith Convolutional Neural Networks”. In: (2015).
- [2] Yingli Tian Chenyang Zhang. “Multi-Modality American Sign LanguageRecognition”. In: (2016).
- [3] Bo Li Qiang Li. “Online Finger Gesture Recognition Using Surface Elec-tromyography Signals”. In: (2013), p. 102.
- [4] Celal Savur. “American Sign Language Recognition System by Using Sur-face EMG Signal”. In: (Feb. 2015), pp. 1-10.
- [5] Ala Shaabana.XTREMIS.url:https://github.com/shibshib/XTREMIS_CAPSTONE.
- [6] Stavros M Panas Vasiliki Kosmidou Leontios Hadjileontiadis. “Evaluationof surface EMG features for the recognition of American Sign Languagegestures”. In: (Aug. 2006).