



We-Sign

An Intermediary for Hearing Impaired Based Communication

- Western Engineering Competition
- January 10, 2021
- **(WINKLE):** Izzy Alfred, Liam Briggs, Natalie Connors, Keon Jin, Waleed Sawan, Edward Zurabov

Presentation Overview

Problem Overview

- Problem Statement
- The Numbers
- Limitations
- Other Solutions

Solution

- Function/Purpose
- Demo
- Hardware
- Image Processing
- User Feedback/Response

Concluding Statements

- Profitability Projections
- Potential Marketing Plan
- Environmental & Social Ramifications
- Applications
- Next Steps

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Problem Statement

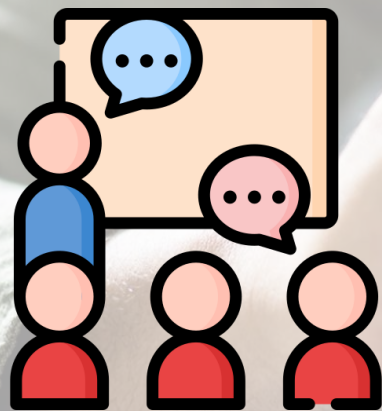
Individuals with auditory disabilities face a universal challenge; their lack of audial clarity imposes natural limitations on them when in pursuit of education, employment, and communication.

Hearing Impairment Statistics

- Roughly 5% of the world's population have hearing loss
- 466 million people have hearing loss greater than 40 dB



The Limitations



Education



Employment



Communication

Other Solutions

Interpreter



Hearing Aid



Cochlear Implant



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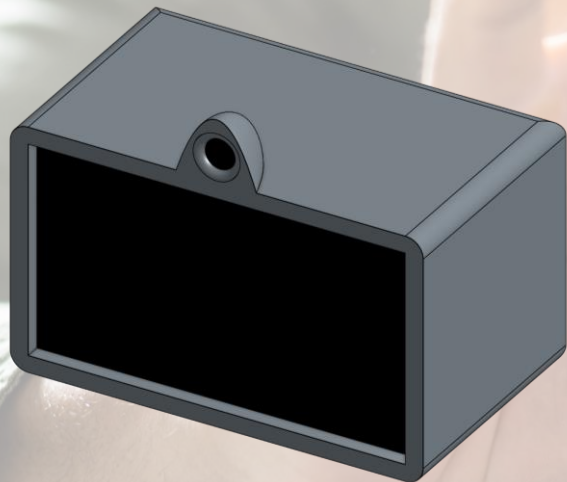
Solution

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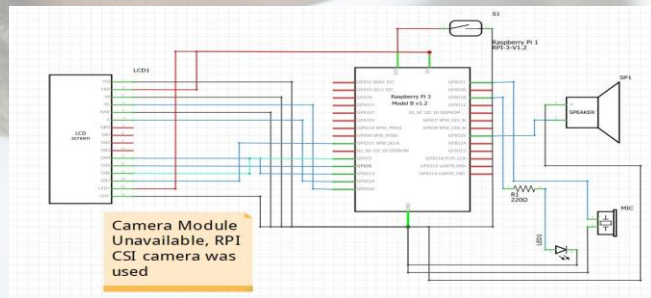
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Function and Purpose



- Users with hearing impairment often face difficulty communicating with others
- Premium equipment for equipment cost incredible sums of money (e.g hearing aids)
- We-sign offers both ASL-to-speech and speech-to-text capabilities for fluid conversation



The diagram illustrates the architecture of the We-Sign system, which is designed for hearing-impaired communication. It is structured into three main stages: Input, Processing, and Response.

Input: This stage receives a hand gesture, represented by a hand with a question mark. A large arrow points down to a hardware schematic of a Raspberry Pi 3 Model B v1.2. The schematic shows connections for an LCD screen, a camera module (noted as an RPi CSI camera), a speaker, and a microphone. A text box states: "Camera Module Unavailable, RPi CSI camera was used".

Processing: The core of the system is a **Convolutional Neural Network**. The diagram shows a stack of layers: 2D Convolution, Normalization, Convolution, Normalization, Convolution, Normalization, Convolution, Normalization, and Convolution. Below this stack, a large arrow points up from a box containing three components: **Image Processing**, **ASL Data Set**, and **Word Inference**.

Response: The final stage produces an **Output**, shown as a clipboard with the text: "tk ASL Interpreter", "User 1 (ASL-to-text) message: hi there where can i find pasta", "User 2 (speech-to-text) message: hello you can find pasta in aisle three", and "User 1 (ASL-to-text) message: thanks". Below the clipboard, two wireless signal icons connect to two boxes: **Text-to-Speech** (using pytsx3) and **Voice Recognition** (using Google Assistant).

Input



Processing

Response

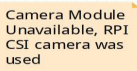



Image Processing

ASL Data Set

Word Inference

Word Inference



pyttsx3

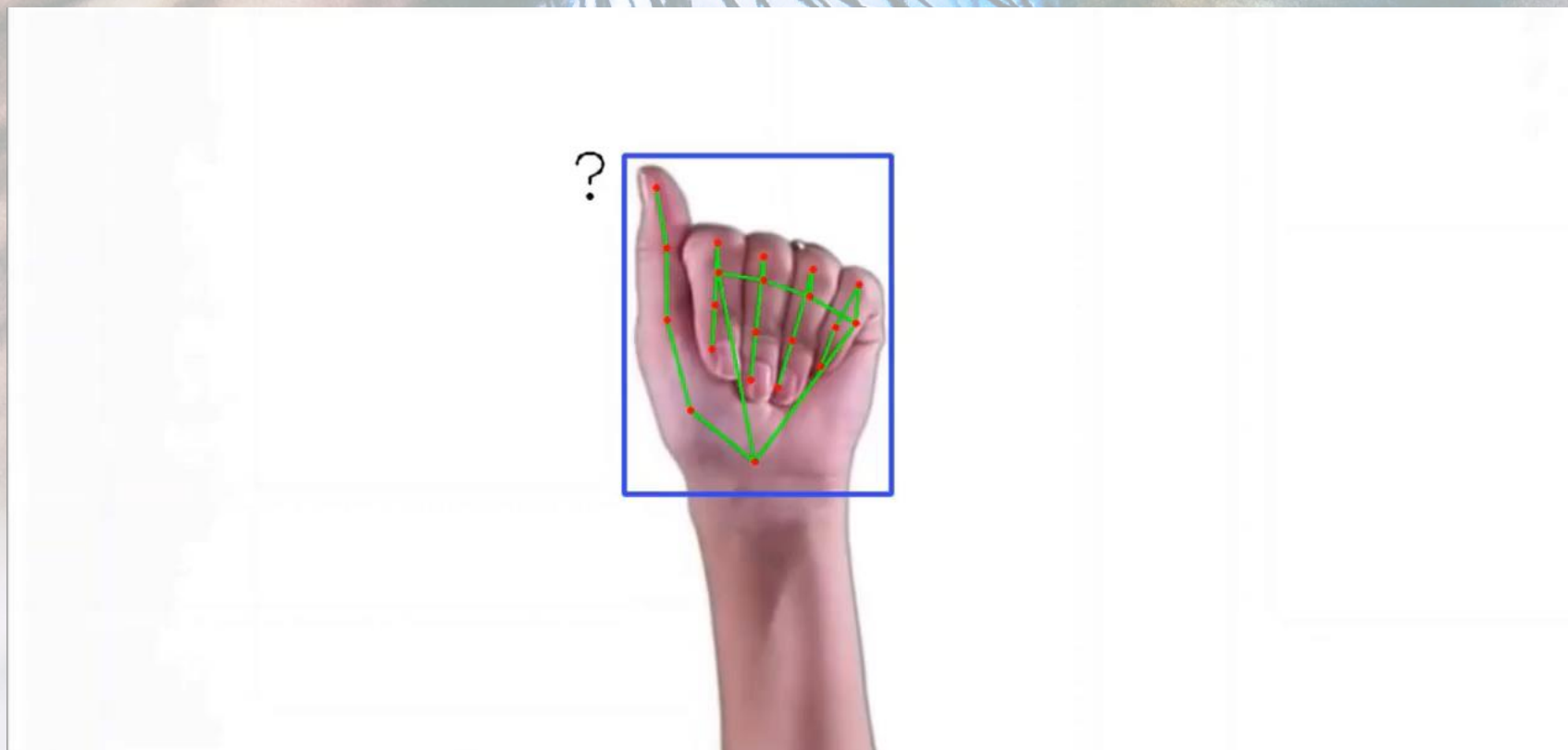
Text-to-Speech



Voice Recognition

Check out the
Video in the
README to see this

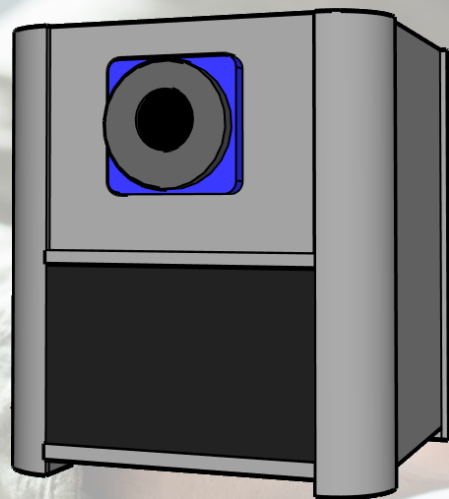
Visual Demonstration



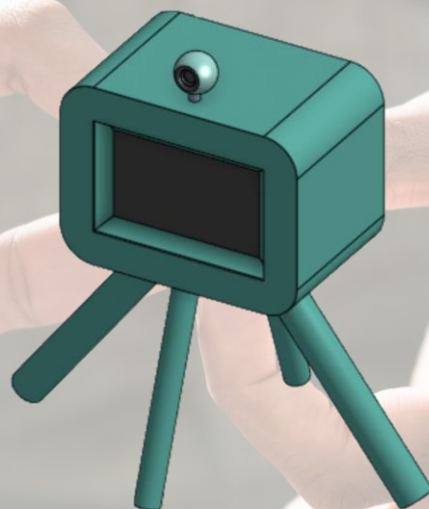
- Hardware
- Image Processing
- User Feedback/Response

Hardware

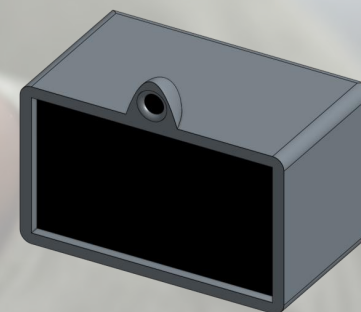
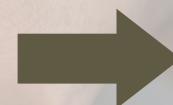
Packaging



Iteration #1



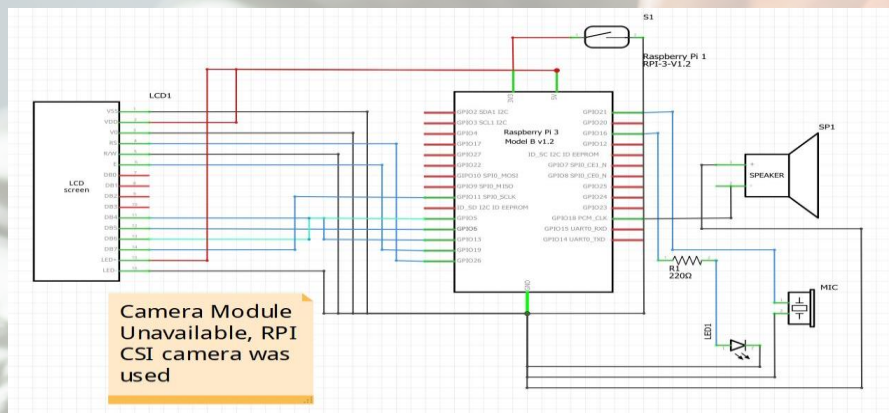
Iteration #2



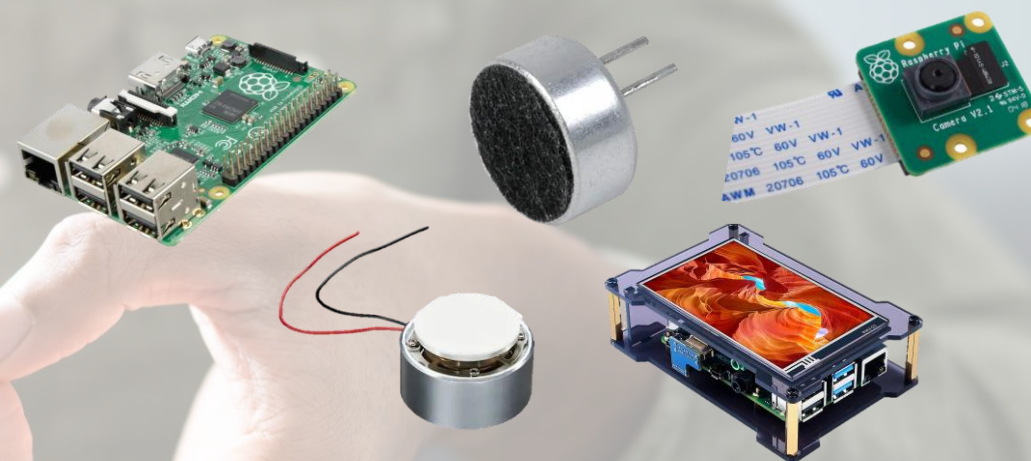
Iteration #3

Hardware

Schematic



Components



Pricing Chart

Item	Cost of Production	
	Prototyping (Source: Amazon)	Manufacturing (Source: Alibaba)
Raspberry Pi	\$99.99	\$31.00
Power Supply Adapter	\$11.85	\$1.02
4-Inch LED Screen	\$44.99	\$23.28
Microphone	\$11.00	\$2.56
Camera	\$39.99	\$8.77
Packaging	\$15.00	\$5.67
Speaker	\$24.71	\$3.50
Total	\$247.53	\$75.80

\$75.80

Prototype Cost: \$247.53

- Hardware
- Image Processing
- User Feedback/Response

Image Processing

Our Approach

Solution

Pre-Classification Image Processing



Neural Network



Feedback

Read in
Input



★ Hand
Location

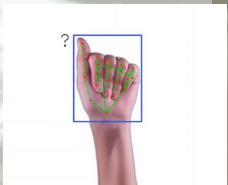
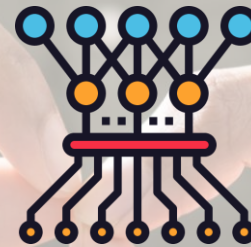


Image Crop
& Grayscale



★ Background
Subtraction



★ Conv. Neural
Network



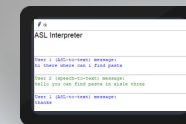
Image Separation



★ Word
Inference



★ Text
to Speech



★ Speech
to Text



Image Processing

Motion-Based Background Subtractor

Our Approach:

- We further advanced the device's image processing capabilities to distinguish hands from its environment using background subtractor techniques
- This was done by extracting the moving foreground (hand) from the static background while minimizing the detection of interference/noise

Our Implementation:

- OpenCV, a library that houses multiple image processing functions, was used to implement background subtraction



Visual Demonstration of
Background Subtraction
Capabilities

Image Processing

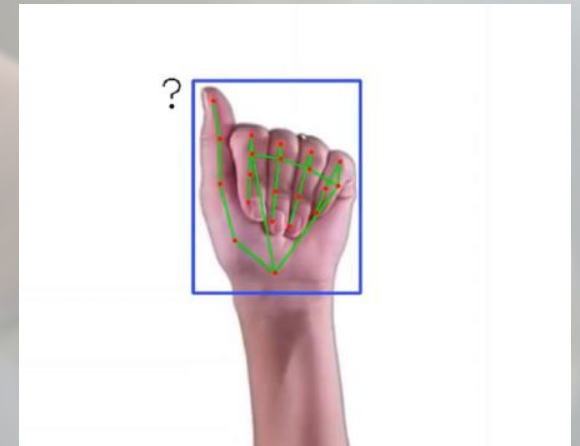
Hand Tracking

Our Approach:

- The need to locate where exactly the hand is in the frame is obviously crucial to ensure that the algorithm is successful
- Locating the hand was done by plotting the location of parts of the hand and using a set to determine a box around

Our Implementation:

- MediaPipe, Google's popular heavily customized ML library, allows us to plot exactly the hand placement
- Additionally, this implementation allows us to expand into more complex phrases in ASL in the future

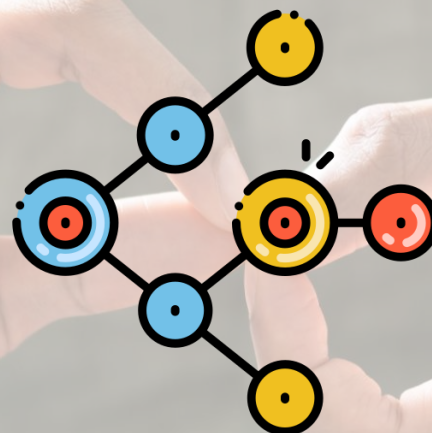


Visual Demonstration of
the ML hand tracking

Conv. Neural Network



**Training Data
(MNIST)**



**Sequential
Layers**



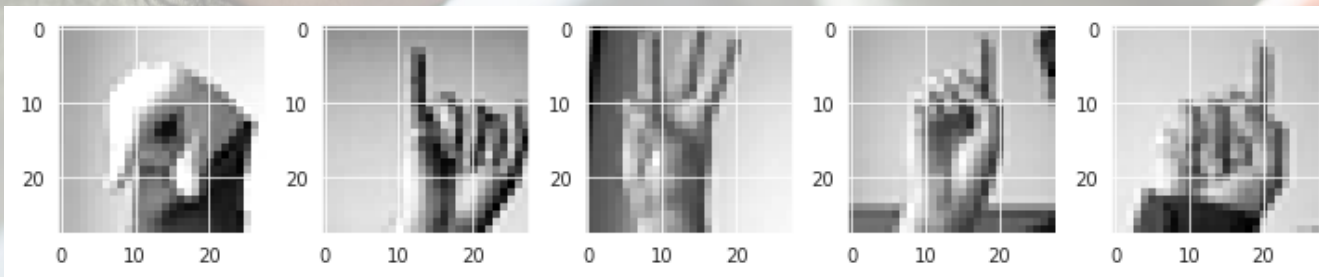
**Model
Accuracy**

Conv. Neural Network

Solution

Training Data

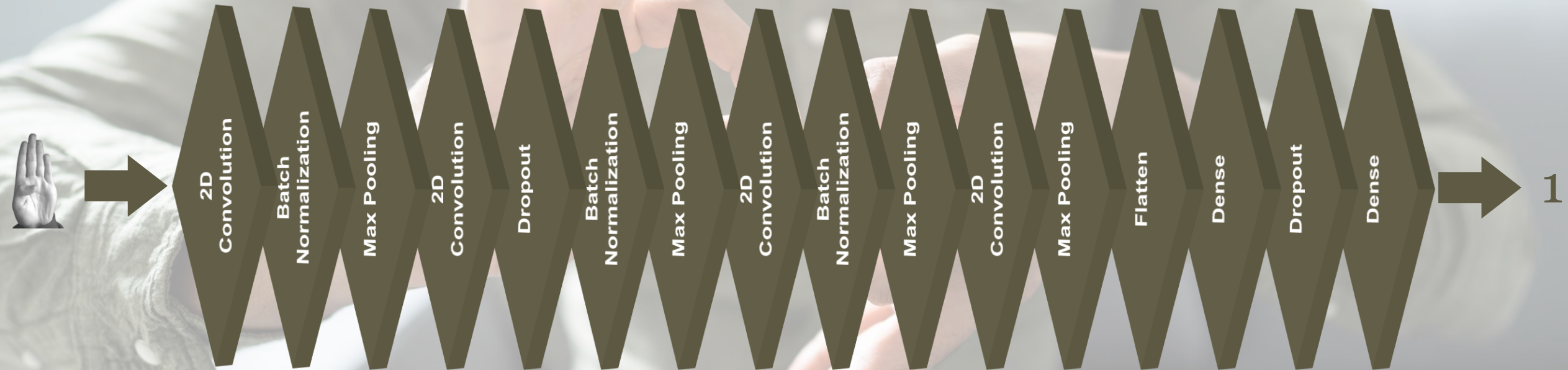
- Used a publicly available ASL training dataset (Sign Language MINST)
- Over 27,000 of training images, 7,000 of testing
- Representing multiple people repeating the standard gestures against different backgrounds



Conv. Neural Network

Solution

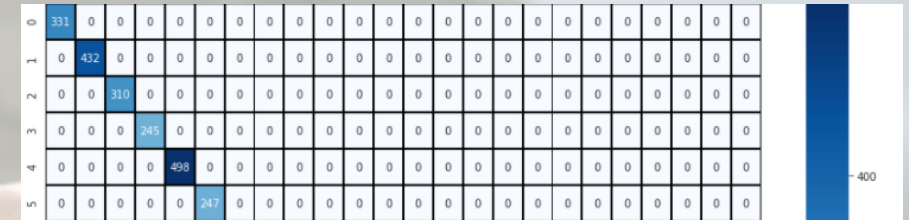
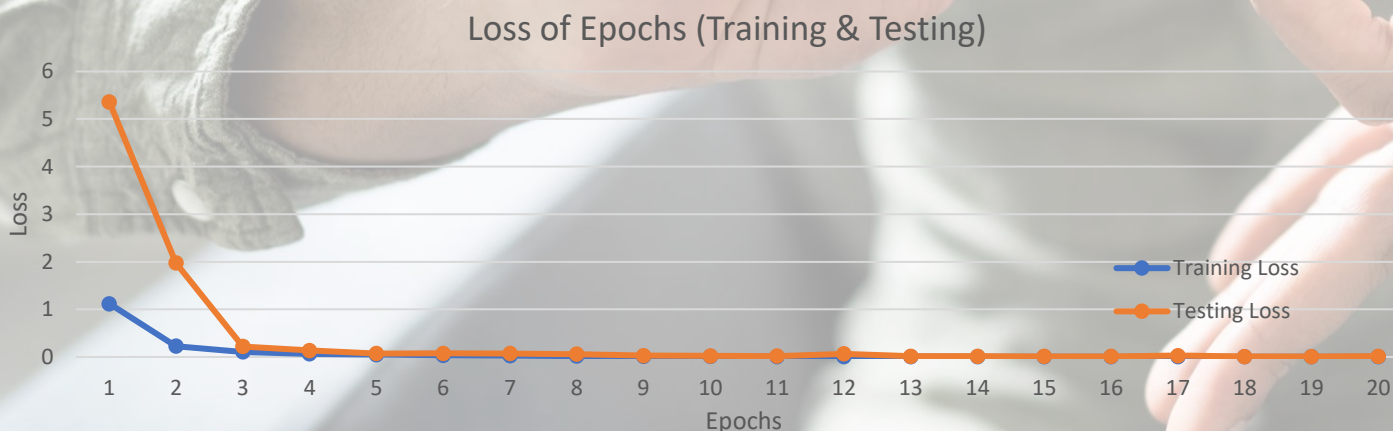
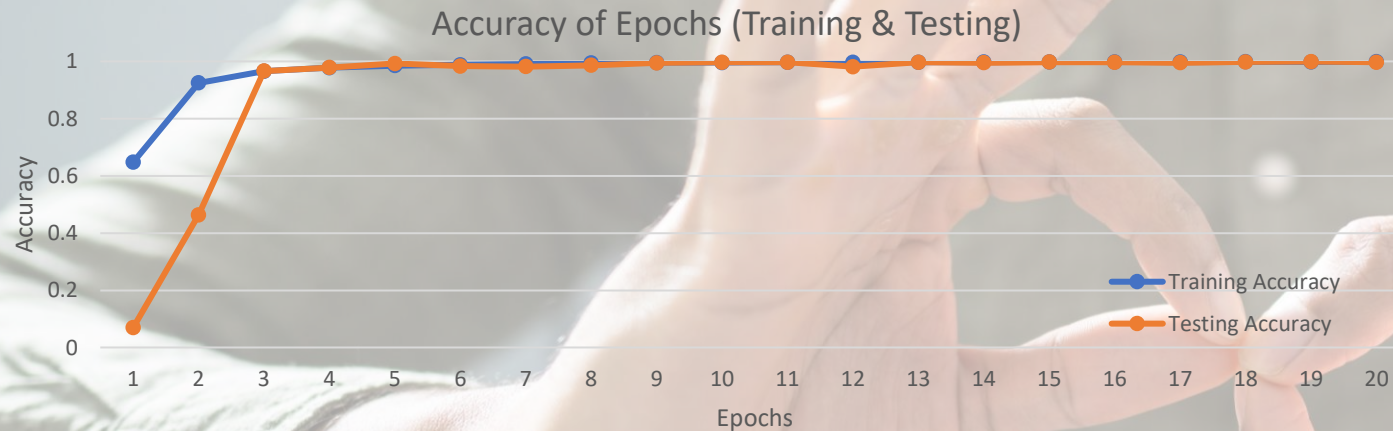
Sequential Layers



Conv. Neural Network

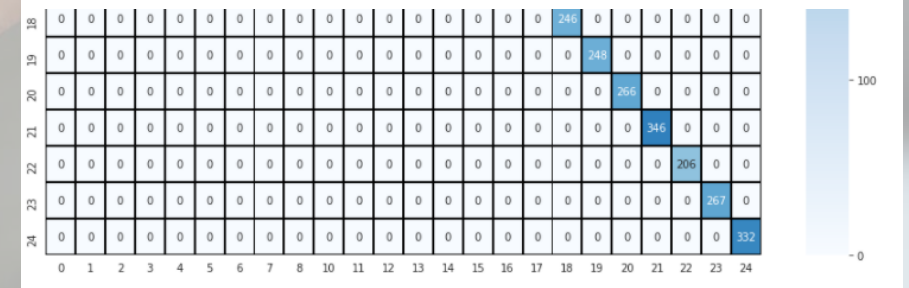
Solution

Model Accuracy



100%

7172/7172
of test cases



- Hardware
- Image Processing
- User Feedback/Response

User Feedback/Response

Word Inference

Character String Transformation

Our Approach:

- The ASL Frame classifier returns a continuous string of characters – string meaning cannot be determined without inferring spaces

Our Implementation:

- Use Zipf's Law – which determines the probability of a word occurrence based on frequency – to construct a cost dictionary of words
- Find lowest-cost string by backtracking cost arrays of different word possibilities

```
1 from math import log
2
3 # Build a cost dictionary, assuming Zipf's law and cost = -math.log(probability).
4 words = open("word-frequency.txt").read().split()
5 wordcost = dict((k, log((i+1)*log(len(words)))) for i,k in enumerate(words))
6 maxword = max(len(x) for x in words)
7
8 def stringToWords(s):
9     # dynamic programming to infer location of spaces in string
10
11     # find best match, given for i characters given i-1 cost
12     # returns a pair (match_cost, match_length).
13     def best_match(i):
14         candidates = enumerate(reversed(cost[max(0, i-maxword):i]))
15         return min((c + wordcost.get(s[i-k-1:i], 9e999), k+1) for k,c in candidates)
16
17     # build the cost array.
18     cost = [0]
19     for i in range(1, len(s)+1):
20         c,k = best_match(i)
21         cost.append(c)
22
23     # backtrack to recover the minimal-cost string.
24     out = []
25     i = len(s)
26     while i>0:
27         c,k = best_match(i)
28         assert c == cost[i]
29         out.append(s[i-k:i])
30         i -= k
31
32     return " ".join(reversed(out))
```

**Word inference from
ASL character string
code component**

Text-to-Speech

Character String Transformation

Our Approach:

- A text-to-speech component from user ASL input encourages seamless and convenient communication between two parties

Our Implementation:

- Pyttsx3 – a text-to-speech Python library – is implemented to express user ASL input in voice-form

pyttsx3 2.90

```
1 import pyttsx3
2
3 # initialize engine
4 engine = pyttsx3.init()
5
6 # set properties
7 engine.setProperty('rate', 150) # speed percent
8 engine.setProperty('volume', 0.9) # Volume 0-1
9
10 # Text-to-speech function
11 def speak(s):
12     engine.say(s)
13     engine.runAndWait()
```

**Text to Speech
Component**

Voice Recognition

Solution

User Interface

Our Approach:

- Main criteria in designing the user interface consisted of enabling a seamless interaction between the two users
- A speech-to-text peripheral enables the device to be efficiently used and avoids interference with ongoing tasks

Our Implementation:

- Using the speech-to-text Google Cloud API, our device can accurately pick up on verbal phrases to be displayed on the screen towards the impaired user



Google Cloud

```
1 import speech_recognition as sr
2
3 r = sr.Recognizer()
4 speech = sr.Microphone(device_index=1)
5 with speech as source:
6     print("say something!")
7     audio = r.adjust_for_ambient_noise(source)
8     audio = r.listen(source)
9
10 try:
11     recog = r.recognize_google(audio, language = 'en-US')
12     print("You said: " + recog)
13 except sr.UnknownValueError:
14     print("Google Speech Recognition could not understand audio")
15 except sr.RequestError as e:
16     print("Could not request results from Google Speech Recognition service; {0}".format(e))
```

Speech to Text
Component

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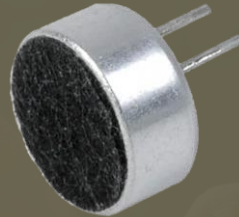
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Cost Breakdown



Key Components

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Manufacturing:
\$75.80/Unit
 Prototyping:
\$247.53/Unit

Projected Profitability

We-Sign Inc. Projected Income Statement For Years Ending 2021 & 2022			
Item	Assumption	2021	2022
Revenue			
Net Sales	2021: Sold 10 Units 2022: Sold 50 Units @ \$300/Unit	\$3,000	\$15,000
COGS	\$75.80/Unit (See Cost Breakdown of Manufacturing)	\$758	\$3,790
Gross Profit		\$2,242	\$11,210
Expenses			
General & Administrative Expense	5% of Net Sales	(150)	(750)
Utilities Expense	3% of Net Sales	(90)	(450)
Advertisement Expense	30% of Net Sales	(900)	(4,500)
Research & Development Expense	10% of Net Sales (Continuous Prototyping)	(300)	(1,500)
Total Expenses		(1,440)	(7,200)
Income Before Tax		\$802	\$4,010
Income Tax	30%	(241)	(1,203)
Net Income		\$561	\$2,807

Key Assumptions:

Unit Price: Approximately manufacturing cost per unit x4

Units Sold: Based on potential partnerships with small-medium institutions

Expenses: Small portions of revenue would be allocated to basic operating expenses, with an additional large focus on advertisement

Income Tax: A conservative yet common rate placed on most corporate entities

Marketing Plan



Environmental & Social Ramifications



Production



Delivery



**End of Useful
Life**

Mitigations

Environmental Mitigations

- Bringing production processes to Canada specifically in locations that use primarily renewable energy, this would also simultaneously lower We-sign's transportation/shipping footprint
- Create a safe disposal program to deal with the hardware after it has reached its useful life

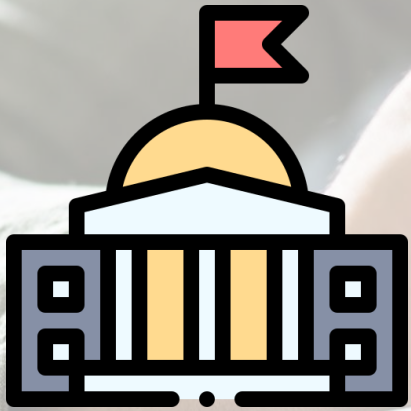


Social Mitigations

- Work with associations such as the Ontario Association of the Deaf to determine those that need the device the most



Applications



**Government
Service Centers**



Retail Stores



Education

Next Steps

Expanded Functionality:



The integration of the ML hand detection algorithm to enables a unique opportunity to increase the functionality to include more mobile phrases

Educational Capabilities:



Implement a learning program alongside the device so that education providers can use the tool to promote ASL learning



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