REAL-TIME COMMUNICATION SYSTEM POWERED BY AI FOR SPECIALLY ABLED

NALAIYA THIRAN PROJECT REPORT

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ABSTRACT:

Artificial Intelligence(AI) has proved its power and efficiency into different domains(industry, medicine, economic, etc), hoping for giving a chance for people with disabilities to change their life for the best. Basically, AI offers many advanced tools and machine learning algorithms that could support human capabilities, needs and preferences and improve User Experience (UX). In this paper, we focus on the impact of AI in Human Machine Interaction. Likewise, we aim to demonstrate what is the impact of AI on disability Human life. Finally, we proceed by introducing a revolutionary approach of adaptive user interfaces to users with special needs.

1.INTRODUCTION

1.1 PROJECT OVERVIEW

Sign language recognition is the process of translating the user's gestures and signs into text. It aids those who are unable to interact with the general population in communication Using image processing methods and neural networks, the motion is mapped to pertinent text in the training data, transforming unprocessed photos and videos into text that can be read and understood. People who are dumb are typically prohibited from having regular conversations with other people in society. They sometimes struggle to communicate with regular people through gestures because the majority of people only recognise a small number of them. People who are deaf or have hearing loss are unable to communicate vocally, so they must frequently use some type of visual communication. The primary form of communication for the deaf and dumb community is sign language. Similar to other languages, it contains grammar and vocabulary, but it communicates primarily through images.

1.2 PURPOSE

The recognition of sign language gestures from real time video and successfully classifying it into either one from a list of categories have been a popular and challenging field of research. Many researchers have been working on this field for a long time, so we have also thought of contributing to this field as by working on it in our final year major project. Liang et al. have also put their research on this concept which has guided us throughout the implementation. The process of recognizing a sign language gesture and classifying it is the single line definition of the task performed by this proposed system. Along with this, a text to ASL finger spelling feature is also available that makes the two-way communication from sign to text and text to sign

possible. The following steps were taken while working on this project. Many vision-based and sensor-based techniques have been used for sign language recognition. Pavlovic et al. The paper published on 1997 emphasizes on the advantages and shortcomings and important differences in the gesture interpretation approaches depending on whether a 3D model of thehuman hand or an image appearance model of the human hand is used. As of the time, this survey was done 3D hand models offered a way of more elaborate modeling of hand gestures but lead to computational hurdles that had not been overcome given the real-time requirements of HCI.

2. LITERATURE REVIEW

S.NO	PAPER	AUTHORS	DESCRIPTION
	TITLE		
1	Portable Communication Aid for Specially Challenged: Conversion of Hand Gestures into Voice and Vice Versa	T Meera Devi, K M Shravan Raju	The goal of the project is to provide a portable communication tool for people with disabilities who have trouble adequately communicating with healthy people. The process of identifying the characteristic hand gesture that sets it apart involves several steps. Using a neural network, the gathered gesticulation is taught. A continuous recording of gestures is isolated from the hand movement pattern. The gestural section represents low-level comprehension of the feature pattern. This will let normal people and those with disabilities communicate with one another.
2	Real-Time Two-Way Communication Approach for Hearing Impaired and Dumb Person Based on Image Processing.	Shweta. S. Shinde, Rajesh M. Autee, Vitthal K. Bhosal	The proposed system employs a method for hand recognition based on vision. The hand movements are recognised in various lighting situations. The suggested method segments the hand's background using the data that has been collected, and then each letter is given a specific gesture. It uses feature extraction techniques to

			determine hand motions' peak and angle values. By translating the motions into speech and vice versa, the gestures are finally recognised. Mel-frequency cepstrum coefficients and dynamic temporal warping are utilised to extract the voice signal's characteristics. The suggested system is MATLAB-based.
3	Full Duplex Communication System for Deaf & Dumb People	Shraddha R. Ghorpade, Surendra K. Waghamare (2015)	People with disabilities are having a difficult time keeping up with the rapidly advancing technology, which is one of the major issues that our society is dealing with. For those with disabilities, having access to communication tools has become crucial. Deaf and stupid individuals typically use sign language for however, they struggle to communicate with people who don't comprehend sign language.
4	Smart communication for differently abled people	R. Bhavani , B. Poornima, M. Surya Bharathi , M. Saraswathi	The majority of the tasks we perform on a daily basis involve speaking and hearing. People who are deaf or dumb find it challenging to communicate with those who do not comprehend sign language or misinterpreters. In this study, we developed a straightforward embedded system-based solution to address this issue. To collect data from the deaf and dumb using sign language, we used a flex sensor. The user of the android-based voice software will speak into it when he or she is deaf and needs to communicate. The programme will then convert this specific speech to text, which will be displayed on LCD. Two Flex sensors are utilised to play whatever messages that Dumb People want to play for the user.

5	Real-Time	Muthu	The real-time sign language	
	Recognition of	Mariappan H,	recognition system is designed	
	Indian Sign	Dr Gomathi V	to identify Indian Sign Language	
	Language		motions (ISL). Sign languages	
			often only use hand gestures and	
			facial emotions. The skin	
			segmentation function of	
			OpenCV is used to identify and	
			track the Regions of Interest	
			(ROI) for the purpose of	
			recognising the indications. By	
			using the fuzzy c-means	
			clustering machine learning	
			algorithm, hand motions are	
			trained and predicted.	
			Applications for gesture	
			recognition include game	
			control, Human-Computer	
			Interaction (HCI), sign language	
			interpretation, and gesture-	
			controlled robots and automated	
			houses. Real-time signs are	
			recognised using the suggested	
			approach. Therefore, it is highly	
			helpful for those with hearing	
			and speech impairments to	
			communicate with other	
			individuals.	

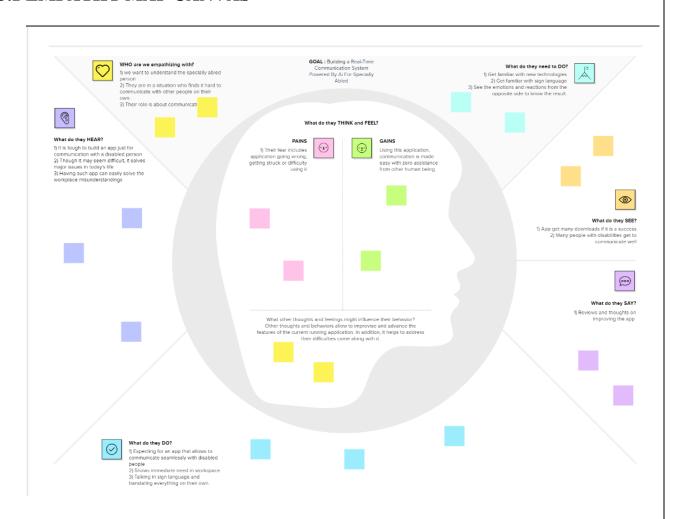
2.1 EXISTING

PROBLEM STATEMENT DEFINITION

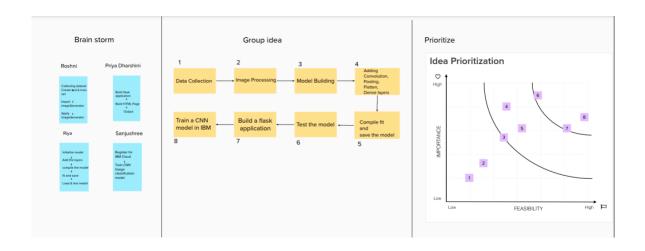
The sign language is used widely by people who are deaf-dumb these are used as a medium for communication. A sign language is nothing but composed of various gestures formed by different shapes of hand, its orientations as well as the facial expressions. There are around 466 million people worldwide with hearing loss and 34 million of these are children. 'Deaf' people have very little or no hearing ability. They use sign language for communication. People use different sign languages in different parts of the world. Compared to spoken languages they are very less in number. In existing system, lack of datasets along with variance in sign language with locality has resulted in restrained efforts in finger gesture detection. Existing project aims at taking the basic step in bridging the communication gap between normal people and deaf and dumb people using Indian sign language. Effective extension of this project to words and common expressions may not only make the deaf and dumb people communicate faster and easier with outer world, but also provide a boost in Developing autonomous systems for understanding and aiding them. The Indian Sign Language lags behind its American Counterpart as the research in this field is hampered by the lack of standard datasets

1. IDEATION & PROPOSED SOLUTION

3.1 EMPATHY MAP CANVAS



3.2 IDEATION & BRAINSTORMING



3.3 PROPOSED SOLUTION

S.NO	PARAMETER	DESCRIPTION
1	Problem Statement (Problem to be solved)	An application for deaf and dumb people to convey their information using signs which get converted to human-understandable language and speech in Artificial Intelligence.
2	Idea / Solution description	 By using voice conversation system with hand gesture recognition and translation will be very useful to havea proper conversation. This makes two people to communicate in a easier and efficient way
3	Novelty / Uniqueness	We are using Convolution neural network to create a model that is trained on different hand gestures andan app is built for the use this mode. • AI has been deployed/used in an efficient manner in the development of this idea.
4	Social Impact / Customer Satisfaction	Communicating with others and being connected in the society andremove accessibility barriers. • With clear results, two people can communicate with each other.
5	Business Model (Revenue Model)	 First, we offer free usage for everyone. Once our product is familiarized among people, we will turn the users to get the premium subscriptions by offering them withmore premium features. As long as our product is beneficial to the users, subscriptions will increase which is a great business

		model, as this worked in case of manytop MNC's around the world.
5	Problem Statement (Problem to be solved)	 Useful in organization where communication between co-workersis much important. Can make collaborations with government. Example: Tata Group made a collaboration with Indian governmentin educating people through Cutting- Edge, which is a great success.

3.4 Problem Fit Solution

4. REQUIREMENT ANALYSIS

FUNCTIONAL REQUIREMENT

Following are the functional requirements of the proposed solution.

FR No.	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)		
FR-1	User Registration	Registration through Form Registration through Gmail		
FR-2	User Confirmation	Confirmation via Email		
FR-3	Image Upload	Upload image through camera Upload image from Gallery		
FR-4	System	Desktop/mobile with good quality camera		
FR-5	Sign Language Conversion	Conversion of sign language into text using Convolutional Neural Network (CNN)		
FR-6	System allowance/access	Provide system access to capture images/videos		

Non-functional Requirements:

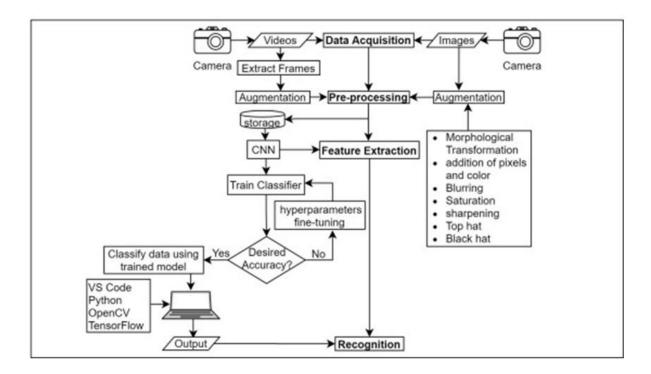
Following are the non-functional requirements of the proposed solution.

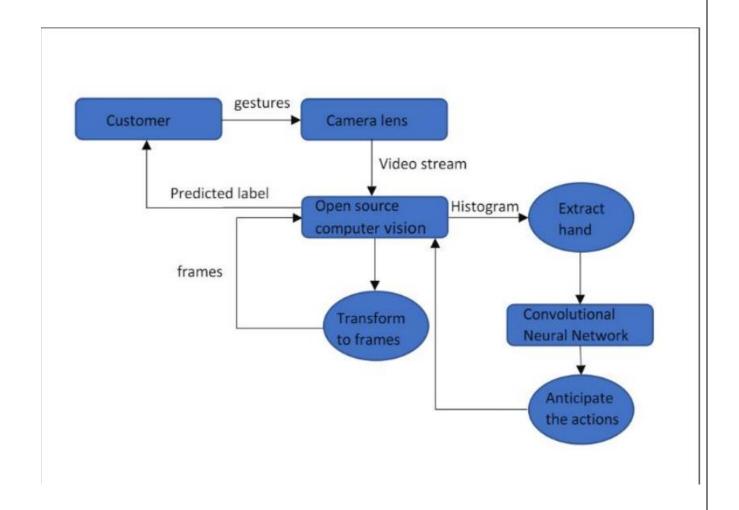
FR	Non-Functional	Description
No.	Requirement	
NFR-	Usability	This system is easy to use.
1		It helps in translating sign language into our desired language.
NFR-	Security	Only the user can sign in to the system and unauthorised users will
2		be avoided at max.
NFR-	Reliability	Accuracy, Predictability to a greater level
3		
NFR-	Performance	After receiving the sign language input from the user, translation is
4		done faster that normal people can understand.
NFR-	Availability	The system is accessible for every user at given point of time.
5		
NFR-	Scalability	Servers helps in achieving scalability whenever needed especially
6		when the number of users increase.

5. PROJECT DESIGN

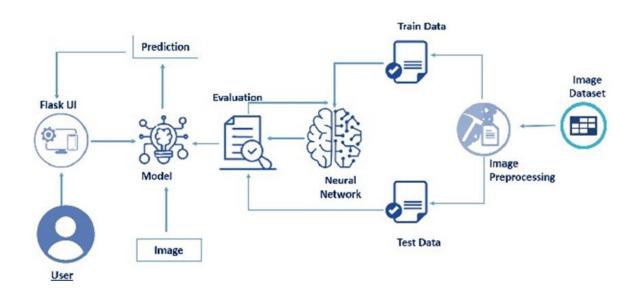
5.1 DATA FLOW DIAGRAMS

A data flow diagram is a two-dimensional diagram that explains how data is processed and transferred in a system. The graphical depiction identifies each source of data and how it interacts with other data sources to reach a common output. Individuals seeking to draft a data flow diagram must identify external inputs and outputs, determine how the inputs and outputs relate to each other, and explain with graphics how these connections relate and what they result in.





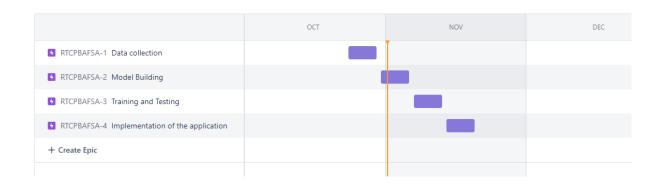
5.2 SOLUTION & TECHNICAL ARCHITECTURE



Software architecture involves the high level structure of software system abstraction, by using decomposition and composition, with architectural style and quality attributes. A software architecture design must conform to the major functionality and performance requirements of the system, as well as satisfy the non-functional requirements such as reliability, scalability, portability, and availability. Software architecture must describe its group of components, their connections, interactions among them and deployment configuration of all components.

PROJECT PLANNING & SCHEDULING

6.1 SPRINT PLANNING & ESTIMATION



6.2 SPRINT DELIVERY SCHEDULE

Project Tracker, Velocity & Burndown Chart: (4 Marks)

Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	10	6 Days	24 Oct 2022	29 Oct 2022	8	29 Oct 2022
Sprint-2	10	6 Days	31 Oct 2022	04 Nov 2022	5	04 Nov 2022
Sprint-3	10	6 Days	07 Nov 2022	11 Nov 2022	7	11 Nov 2022
Sprint-4	10	6 Days	14 Nov 2022	18 Nov 2022	5	18 Nov 2022

6.3 REPORTS FROM JIRA

JIRA has categorized reports in four levels, which are

- Agile
- Issue Analysis
- Forecast & Management
- Others

Velocity:

$$AV = \frac{sprint\ duration}{velocity}$$

$$AV = 6/10 = 0.6$$

Burndown chart:



SPRINT BURNDOWN



7. CODING & SOLUTION

7.1 FEATURE 1

from tensorflow.keras.preprocessing.image **import** ImageDataGenerator **as** ig train = ig(rescale = 1./255, shear_range=0.2, zoom_range=0.2, horizontal_flip=**True**) test = ig(rescale=1./255)

xtrain = train.flow_from_directory(r'E:\ibm\IBM-Project-31230-1660197974\Data Collection\Dataset\training_set', target_size=(64,64), batch_size=300, class_mode='categorical', color_mode= 'grayscale')

ytrain = train.flow_from_directory(r'E:\ibm\IBM-Project-31230-1660197974\Data Collection\Dataset\test_set', target_size=(64,64), batch_size=300, class_mode='categorical', color_mode= 'grayscale')

Found 15750 images belonging to 9 classes.

Found 2250 images belonging to 9 classes.

```
In [5]:

from tense file display train = ig. File display 1./255, shear_range=0.2, zoom_range=0.2, horizontal_flip=True) test = ig(rescale=1./255)

In [7]: xtrain = train.flow_from_directory(r'E:\ibm\IBM-Project-31230-1660197974\Data Collection\Dataset\training_set', target_size=(64,64), batch_size=300, c ytrain = train.flow_from_directory(r'E:\ibm\IBM-Project-31230-1660197974\Data Collection\Dataset\test_set', target_size=(64,64), batch_size=300, class Found 15750 images belonging to 9 classes.

Found 2250 images belonging to 9 classes.
```

MODEL BUILDING

```
model = Sequential()
model.add(Convolution2D(32, (3,3), input_shape=(64,64,1), activation='relu')) #convolution layer
model.add(MaxPooling2D(pool_size=(2,2))) # pooling layer
model.add(Flatten()) #flatten layer
model.add(Dense(units=512, activation='relu')) #dense layer
model.add(Dense(units=9, activation='softmax')) #dense layer
```

```
model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
                                                                                                         In [20]:
      model.fit_generator(xtrain, steps_per_epoch=24, epochs=10, validation_data=xtest, validation_steps=40) #fit the model
      model.save('aslpng1.h5') #saving the model
       C:\Users\riyap\anaconda3\envs\tf-gpu\lib\site-packages\tensorflow\python\keras\engine\training.py:1940: UserWarning: `Model.fit_generator
       d and will be removed in a future version. Please use `Model.fit`, which supports generators.
        warnings.warn('`Model.fit_generator` is deprecated and
       Epoch 1/10
       24/24 [============] - ETA: 0s - loss: 1.0995 - accuracy: 0.6942WARNING:tensorflow:Your input ran out of data; interrupting t
       Make sure that your dataset or generator can generate at least `steps_per_epoch * epochs` batches (in this case, 40 batches). You may need to
       epeat() function when building your dataset.
       Epoch 2/10
       24/24 [================== ] - 22s 916ms/step - loss: 0.2447 - accuracy: 0.9319
       Epoch 3/10
       Epoch 4/10
       Epoch 5/10
               24/24 [====
       Epoch 6/10
       24/24 [============= ] - 8s 311ms/step - loss: 0.0422 - accuracy: 0.9893
       Epoch 7/10
       Epoch 8/10
       Epoch 9/10
       Epoch 10/10
       TESTING MODEL:
      from tensorflow.keras.models import load_model
      import numpy as np
      import cv2
      import skimage
      model = load_model('aslpng1.h5')
      #preprocess the images
      from skimage.transform import resize
      def detect(frame):
        img = resize(frame, (64,64,1))
        img =np.expand dims(img, axis=0)
        if(np.max(img)>1):
          img = img/255.0
        pred = model.predict(img)
        print(pred)
        pred = model.predict_classes(img)
        print(pred)
      #Load and predict
      frame = cv2.imread(r'E:\ibm\IBM-Project-31230-1660197974\Data Collection\Dataset\test_set\G\1.png')
      data = detect(frame)
[[4.5008630e-08 5.9901827e-11 1.7369990e-06 3.4444175e-10 1.1011272e-09
 7.6371736e-09 9.9998164e-01 1.6525357e-05 1.1666364e-10]]
C:\Users\riyap\anaconda3\envs\tf-gpu\lib\site-packages\tensorflow\python\keras\engine\sequential.py:455: UserWarning: `model.predict_classes()`
ecated and will be removed after 2021-01-01. Please use instead:* `np.argmax(model.predict(x), axis=-1)`,
                                                                               if your model does multi-class classificati
    e.g. if it uses a `softmax` last-layer activation).* `(model.predict(x) > 0.5).astype("int32")`,
                                                                             if your model does binary classification
g. if it uses a `sigmoid` last-layer activation).
```

[6]

warnings.warn('`model.predict_classes()` is deprecated and '

8. TESTING

TEST CASES

A test case has components that describe input, action and an expected response, in order to determine if a feature of an application is working correctly. A test case is a set of instructions on "HOW" to validate a particular test objective/target, which when followed will tell us if the expected behavior of the system is satisfied or not.

Characteristics of a good test case:

- 3.5.1 Accurate: Exacts the purpose.
- 3.5.2 Economical: No unnecessary steps or words.
- 3.5.3 Traceable: Capable of being traced to requirements.
- 3.5.4 Repeatable: Can be used to perform the test over and over.
- 3.5.5 Reusable: Can be reused if necessary

S.N	FUNCTION	DESCRIPTIO	EXPECTE	ACTUA	STATU
O		N	DOUTPUT	L	$ \mathbf{S} $
				OUTPU T	
1	Framework	Generate the	Individual	Individual	Success
	construction	GUI for admin	page for	page for	
		and user	admin and	admin	
			user	and user	
2	Read the	Comments	Comments	Comment	Success
	comments	Analysis	in text	s in text	
			format	format	
3	Classification	Classify the	Finger	Finger	Success
		Datasets	Gestures	Gestures	
4	Rules	Block the	Block the	Block the	Success
	implementatio	comments and	users	users	
	n	friends			

3.6 USER ACCEPTANCE TESTING

Acceptance testing can be defined in many ways, but a simple definition is the succeeds when the software functions in a manner that can be reasonable expected by the customer. After the acceptance test has been conducted, one of the two possible conditions exists. This is to fine whether the inputs are accepted by the database or other validations. For example accept only numbers in the numeric field, date format data in the date field. Also the null check for the not null fields. If any error occurs then show the error messages. The function of performance characteristics to specification and is accepted. A deviation from specification is uncovered and a deficiency list is created. User Acceptance Testing is a critical phase of any project and requires significant participation by the end user. It also ensures that the system meets the functional requirements.



Realtime communication system powered by A.I. for specially abled

A sign language is a way of communicating by using the hands and other parts of the body. It should not be confused with body language. Sign languages are an important way for deaf people to communicate.

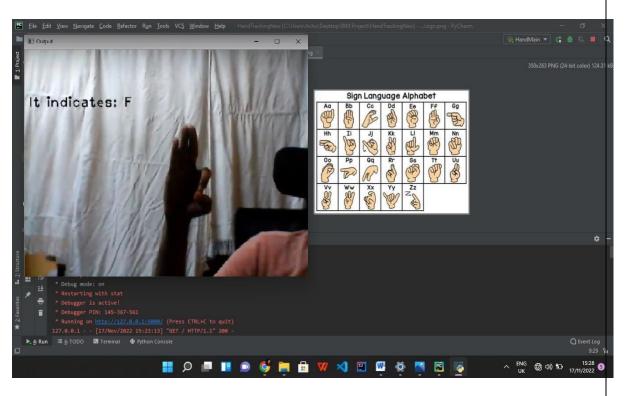


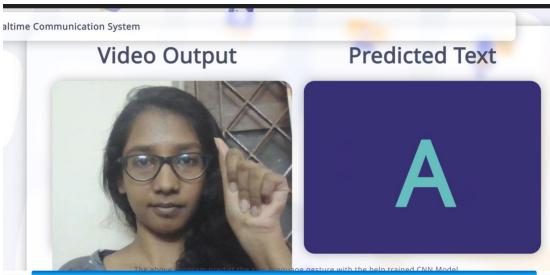
START TRANSLATE

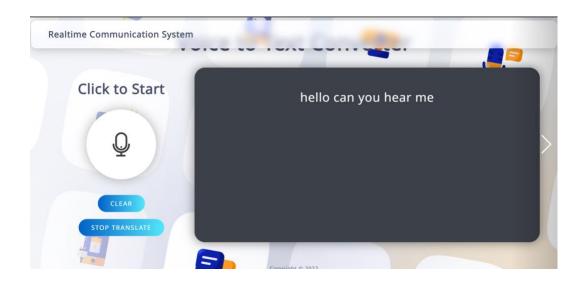


9. RESULTS

9.1 PERFORMANCE METRICS







10. ADVANTAGES & DISADVANTAGES

DISADVANTAGES

- i. Need hardware control to detect the hands
- ii. Hand segmentation become complex of various backgrounds
- iii. Segmentation accuracy is less in hand tracking

ADVANTAGES

- Segmentation accuracy is high
- Easy to detect the finger postures
- Track fingers and sign recognition with less computational steps
 No need for additional hardware system

11. CONCLUSION

The ability to look, listen, talk, and respond appropriately to events is one of the most valuable gifts a human being can have. However, some unfortunate people are denied this opportunity. People get to know one another through sharing their ideas, thoughts, and experiences with others around them. There are several ways to accomplish this, the best of which is the gift of "Speech." Everyone can very persuasively transfer their thoughts and comprehend each other through speech. Our initiative intends to close the gap by including a lowcost computer into the communication chain, allowing sign language to be captured, recognised, and translated into speech for the benefit of blind individuals. An image processing technique is employed in this paper to recognise the handmade movements. This application is used to present a modern integrated planned system for hear impaired people. The camera-based zone of interest can aid in the user's data collection. Each action will be significant in its own right.

12. FUTURE SCOPE

Despite it having average accuracy, our system is still well-matched with the existing systems, given that it can perform recognition at the given accuracy with larger vocabularies and without an aid such as gloves or hand markings. In future, we can extend the framework to implement various deep learning algorithms to recognize the signs and implement in real time applications.

13. APPENDIX

SOURCE CODE

```
from flask import Flask, render_template, flash, request, session from flask import render_template,
       redirect, url_for, request
      import smtplib
                            app = Flask(__name__) app.config.from_object(__name__)
app.config['SECRET_KEY'] = '7d441f27d441f27567d441f2b6176a'
     app.config['DEBUG']
       @app.route("/") def homepage():
         return render template('index.html')
       @app.route("/UserLogin") def UserLogin():
         return render_template('UserLogin.html')
@app.route("/start", methods=['GET', 'POST']) def start():
                        if request.method == 'POST':
         error = None
           import csv
                          import copy
                                           import cv2 as
                                          from model
       cv
              import mediapipe as mp
       import KeyPointClassifier
                                     from app files
       import calc landmark list, draw info text,
       draw_landmarks, get_args, pre_process_landmark
      from PIL import Image, ImageDraw, ImageFont
```

```
args = get_args() cap_device = args.device
cap_width = args.width cap_height = args.height
```

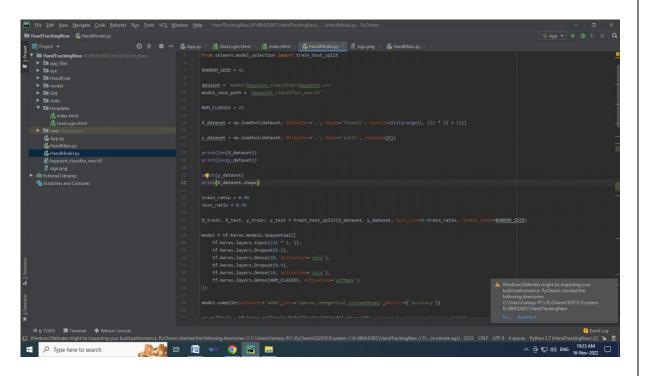
import numpy as np

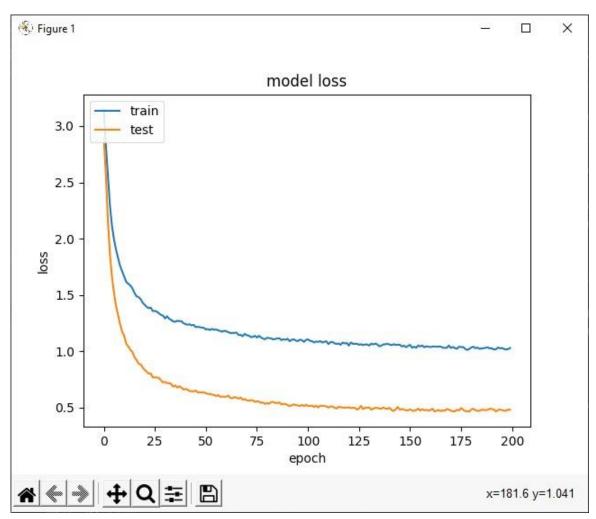
```
use_static_image_mode = args.use_static_image_mode
                                                            min_detection_confidence =
args.min_detection_confidence
                                 min_tracking_confidence = args.min_tracking_confidence
    cap = cv.VideoCapture(cap_device)
                                          cap.set(cv.CAP_PROP_FRAME_WIDTH, cap_width)
cap.set(cv.CAP PROP FRAME HEIGHT, cap height)
    mp_hands = mp.solutions.hands
                                       hands = mp_hands.Hands(
static_image_mode=use_static_image_mode,
                                                max_num_hands=1,
min_detection_confidence=min_detection_confidence,
min_tracking_confidence=min_tracking_confidence,
    keypoint_classifier = KeyPointClassifier()
    with
open('model/keypoint_classifier/keypoint_classifier_label.csv', encoding='utf-8-sig') as f:
      keypoint_classifier_labels = csv.reader(f)
                                                    keypoint_classifier_labels = [
row[0] for row in keypoint_classifier_labels
      1
    flag = 0
    import win32com.client as wincl
                                       speak =
wincl.Dispatch("SAPI.SpVoice")
    while True:
                                 if key == 27: # ESC
      key = cv.waitKey(10)
break
      ret, image = cap.read()
                                   if not ret:
                                                     break
                                                                 image =
cv.flip(image, 1)
                      debug_image = copy.deepcopy(image)
      # print(debug_image.shape)
      # cv.imshow("debug_image",debug_image)
      image = cv.cvtColor(image, cv.COLOR_BGR2RGB)
```

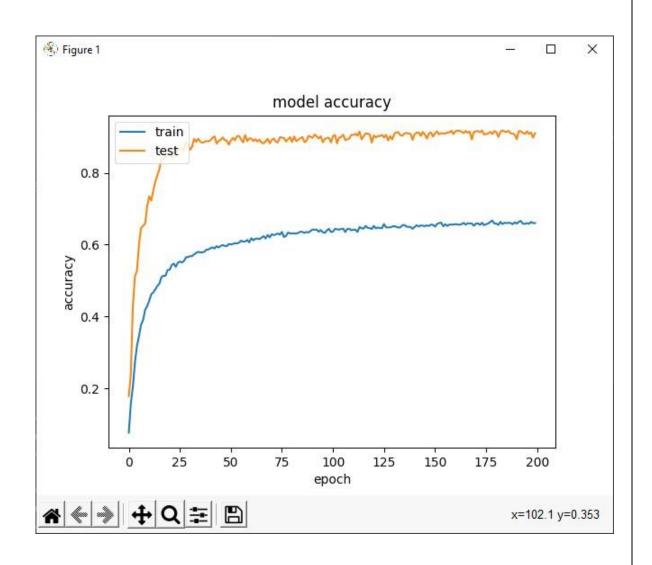
```
image.flags.writeable = False
                                              results = hands.process(image)
     image.flags.writeable = True
           if results.multi hand landmarks is not None:
              for hand landmarks, handedness in zip(results.multi hand landmarks, results.multi handedness):
                landmark list = calc landmark list(debug image, hand landmarks)
                # print(hand_landmarks)
                pre_processed_landmark_list = pre_process_landmark(landmark_list)
                hand sign id =
     keypoint_classifier(pre_processed_landmark_list)
                debug_image = draw_landmarks(debug_image, landmark_list)
                flag += 1
                                    print(flag)
                                                         if
     (flag == 100):
                  flag = 0
                  speak.Speak(keypoint_classifier_labels[hand_sign_id])
                debug_image = draw_info_text(
                  debug_image,
                                             handedness,
     keypoint_classifier_labels[hand_sign_id])
           cv.imshow('Hand Gesture Recognition', debug_image)
         cap.release()
                          cv.destroyAllWindows
                                                        return
     render_template('UserLogin.html')
     if __name__ == '__main__':
app.run(debug=True, use_reloader=True)
     import mediapipe as mp import cv2 import numpy
     as np import uuid import os
```

"import subprocess as sp programName = "notepad.exe"

```
#fileName = "sms.txt"
#sp.Popen([programName, fileName]) sp.Popen([programName])"
mp drawing = mp.solutions.drawing utils mp hands = mp.solutions.hands
cap = cv2.VideoCapture(0)
with mp_hands.Hands(min_detection_confidence=0.8, min_tracking_confidence=0.5) as
hands:
         while cap.isOpened():
                                  ret, frame = cap.read()
    # BGR 2 RGB
                      image = cv2.cvtColor(frame, cv2.COLOR_BGR2RGB)
    # Flip on horizontal
                           image = cv2.flip(image, 1)
    # Set flag
                 image.flags.writeable = False
    # Detections
                    results = hands.process(image)
    # Set flag to true
                        image.flags.writeable = True
    # RGB 2 BGR
                      image = cv2.cvtColor(image, cv2.COLOR_RGB2BGR)
   # print(results)
    # Rendering results
                           if results.multi_hand_landmarks:
                                                                  for num, hand in
enumerate(results.multi_hand_landmarks):
                                                  mp_drawing.draw_landmarks(image, hand,
mp_hands.HAND_CONNECTIONS,
                       mp_drawing.DrawingSpec(color=(14, 22,
76), thickness=2, circle_radius=4),
                                                       mp_drawing.DrawingSpec(color=(24, 44, 250),
thickness=2, circle_radius=2),
                       )
    cv2.imshow('Hand Tracking', image)
    if cv2.waitKey(10) & 0xFF == ord('q'):
                                               break
cap.release() cv2.destroyAllWindows()
```







GITHUB ACCOUNT LINK:

https://github.com/IBM-EPBL/IBM-Project-31230-1660197974

PROJECT DEMO LINK:

