Final Report

A Real Time Communication System For Specially Abled

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1. INTRODUCTION

1.1 Project Overview:

People with impairments are a part of our society. Though technology is constantly evolving, little is being done to improve the lives of these people. It has never been easy to communicate with a deaf-mute individual. It is quite challenging for silent persons to communicate with non-mute people, mostly because hand sign language is not taught to the general public. It might be quite challenging for them to communicate at times of crisis.In circumstances where other modes of communication, like speech, cannot be employed, the human hand has continued to be a preferred choice. A proper discussion between a normal person and an impaired person in any language will be made possible by a voice conversion system with hand gesture recognition and translation. The project intends to create a system that can translate speech into acceptable sign language for the deaf and dumb as well as translate sign language into a human hearing voice in the desired language to communicate a message to normal people. A convolution neural network is being used to build a model that is trained on various hand motions. On the basis of this model, an app is created. With the help of this app, persons who are deaf or dumb can communicate using signs that are translated into speech and human-understandable words.

1.2 Purpose:

The only means of communication for someone who cannot hear or speak is through sign language. People who are physically disabled can communicate their thoughts and feelings more effectively by using sign language. The alphabets and motions of sign language have been identified in this work using a revolutionary system of sign language recognition. We can identify the indications and provide the appropriate text output with the use of computer vision and neural networks.

2. LITERATURE SURVEY

2.1 Existing Problem

In our society, we have people with disabilities. The technology is developing day by day but no significant developments are undertaken for the betterment of these people. Communications between deaf-mute and a normal person has always been a challenging task. It is complicated for mute people to convey their message to normal people. Since normal people are not trained on hand sign language. In emergency times conveying their message is very difficult. The human hand has remained a popular choice to convey information in situations where other forms like speech cannot be used. Voice Conversion System with Hand Gesture Recognition and translation will be handy to have a proper conversation between a normal person and an impaired person in any language.

2.2 References

Design of Communication Interpreter for Deaf and Dumb Person was published by Pallavi Verma (Electrical and Electronics Department, Amity University, Greater Noida, Uttar Pradesh, India), Shimi S. L (Assistant Professor, NITTTR, Chandigarh, India), Richa Priyadarshani (Electrical and Electronics Department, Amity University, Greater Noida, Uttar Pradesh, India).International Journal of Science and Research (IJSR) · Jan 2013

Development of full duplex intelligent communication system for deaf and dumb people was published in the year January 2017 DOI:10.1109/CONFLUENCE.2017.7943247

At 7th International Conference on Cloud Computing, Data Science & Engineering - Confluence (Confluence) by Surbhi Rathi Department of Information Technology, Yeshwantrao Chavan College of Engineering Nagpur, India and Ujwalla Gawande, Department of Information Technology Yeshwantrao Chavan College of Engineering Nagpur, India.

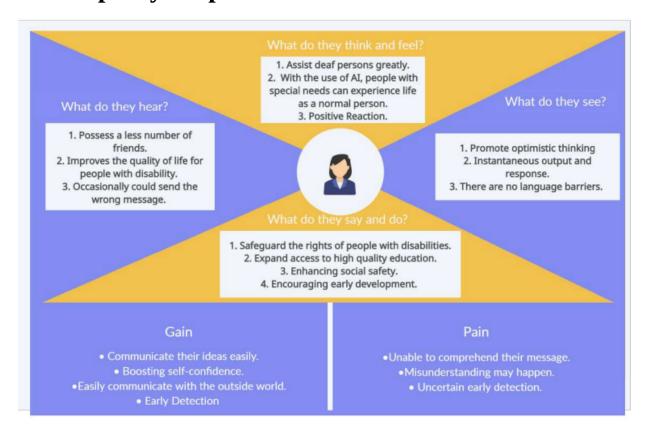
A Review Paper on Sign Language Recognition for The Deaf and Dumb published by R Rumana(B.E Graduate(IV year), Department of Computer Science and Engineering, SCSVMV, Kanchipuram), Reddygari Sandhya Rani(B.E Graduate(IV year), Department of Computer Science and Engineering, SCSVMV, Kanchipuram), Mrs. R. Prema(Assistant Professor, Department of Computer Science and Engineering, SCSVMV, Kanchipuram). Published (First Online): 01-11-2021

2.3 Problem Definition Statement

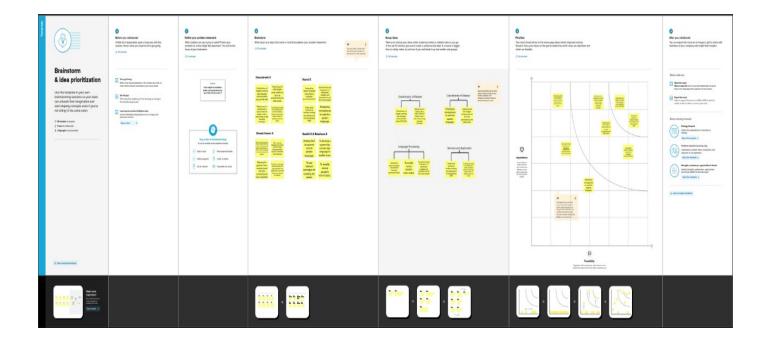
People with disabilities are a part of our society. Even though technology is constantly evolving, little is being done to improve the lives of these people. Communication with a deaf-mute person has always been difficult. Because hand sign language is not taught to the general public, it can be difficult for silent people to communicate with non-mute people. In times of crisis, they may find it difficult to communicate. When other modes of communication, such as speech, are unavailable, the human hand has remained a popular method of information transmission. A voice conversion system with hand gesture recognition and translation will be very helpful in establishing proper communication between a normal person and a handicapped person in any language.

3. IDEATION & PROPOSED SOLUTION

3.1 Empathy Map Canvas



3.2 Ideation & Brainstorming

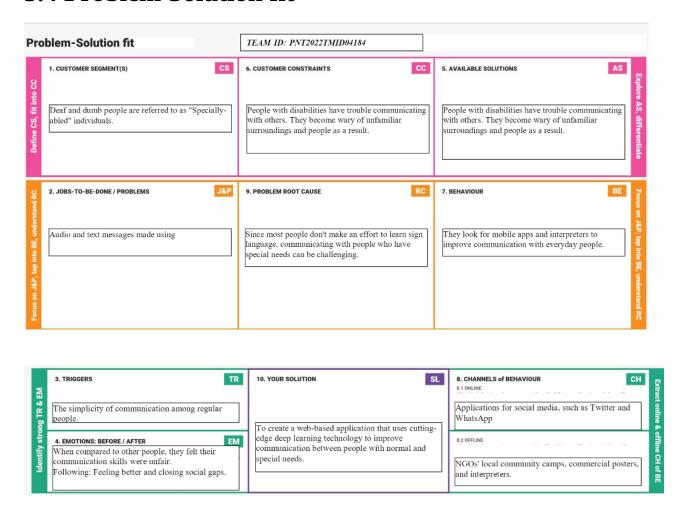


3.3 Proposed Solution

The project's goal is to develop a system that will help deaf and dumb people communicate in real time. The model uses background-subtraction, a method that is thought to be a crucial pre-processing step. These methods will be implemented in the development of the proposed Sign Language Converter in order to locate the hand region in the input image captured by the camera.

We'll be using the most latest and well-liked wearable technology, which allows the disabled person to effortlessly take the gadget (a mobile application) wherever and everywhere, facilitating communication for both people with special needs and regular people. The performance of the trained model will be improved by using the most recent convolution neural network architecture. The goal of our project is to close communication barriers with people who have speech and hearing impairments.

3.4 Problem Solution fit



4 REQUIREMENT ANALYSIS

4.1 Functional Requirements

Following are the functional requirements of the proposed solution.

FR	Functional	Sub Requirement (Story / Sub-Task)
No.	Requirement (Epic)	
FR-1	User Registration	Registration through Form
		Registration through
		Gmail Registration
		through LinkedIN
FR-2	User Confirmation	Confirmation via Email
		Confirmation via OTP
FR-3	User Interface	User friendly and can be handled easily.
FR-4	Collection of datasets	Information is gathered regarding
		the various signs and the
		corresponding voice.

4.2 Non-functional Requirements:

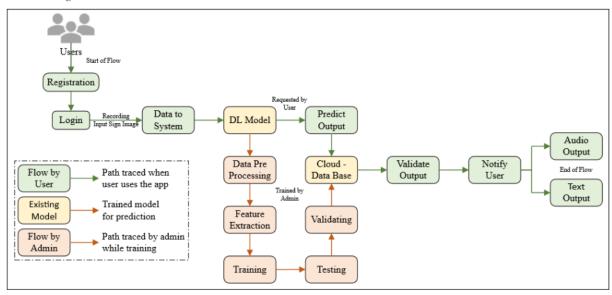
Following are the non-functional requirements of the proposed solution.

FR No.	Non-Functional	Description
	Requirement	
NFR-1	Usability	Used to allow people with special
		needs to speak openly with people
		without any obstacles.
NFR-2	Security	The user of this website, their
		information, and their personal data
		are secured.
NFR-3	Reliability	After correctly determining the exact
		sign conversion for the given sign, their corresponding voice is then
		outputted with high accuracy.
NFR-4	Performance	The proposed approach is more effective
		and can be accessed by manyindividuals
		because it will be made available online
NFR-5	Availability	and has an easy-to-use interface. The website needs to be responsive in
		order to be
		usable on all types of devices, including
		computers, tablets, and mobile phones.
NFR-6	Scalability	The proposed model can be scaled up to
		become a more complex system that can
		recognise several sign languages and also
		translate into multiple
		conventional languages with enough cash
		and labor.

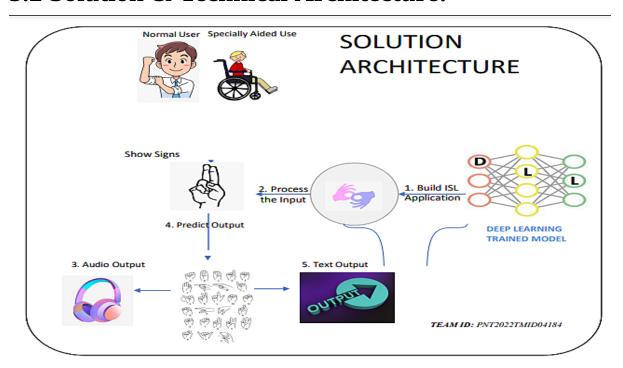
5 PROJECT DESIGN

5.1 Data Flow Diagram:

Data Flow Diagram:



5.2 Solution & Technical Architecture:



5.3 User Stories

User Stories:

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer (Mobile user)	Registration	USN-1	I can sign up for the application as a user by providing my email address, a password, and a password confirmation.	I can access my account / dashboard	High	Sprint - 1
	Authentication	USN-2	After registering for the application, I as a user will receive a confirmation email.	I can receive confirmation email & click confirm	Low	Sprint - 1
	Login	USN-3	I can access the application as a user by providing my email address and password.	I am able to get into the Dashboard	High	Sprint - 2
	Dashboard	USN-4	One place to explore all available features.	I can access my dashboard	High	Sprint - 2
Customer (Web user)	Registration	USN-1	I can sign up for the application as a user by providing my email address, a password, and a password confirmation.	I can access my account / dashboard	Hìgh	Sprint - 1
	Authentication	USN-2	After registering for the application, I as a user will get a confirmation email.	I can receive confirmation email & click confirm.	Low	Sprint - 1
	Login	USN-3	By providing my email address and password, I can access the application as a user.	I am able to get into the Dashboard	Low	Sprint - 2
	Dashboard	USN-4	A single location to discover all features offered	I can access my dashboard	Low	Sprint - 2
	Upload image	USN-5	I can post the sign language image as a user to have it converted to text.	I can be able to see the appropriate text for the sign language	High	Sprint - 3
Administrator	Manage	USN-6	Do-it-yourself service for delivering Everything.	A group of conditions that must be satisfied for a user story to be deemed complete.	High	Sprint - 4

6 PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning an

Product Backlog, Sprint Schedule, and Estimation (4 Marks)

Sprint	Functional Requireme nt (Epic)		User Story / Task	Story Points	Priority	Team Members
Sprint-1	Data Collection	USN-1	Collect Dataset .	9	High	DWARAKNATH K, SWATHI K, HARINI K, DINESH KUMAR S, RAKSHANA S
Sprint-1		USN-2	Image preprocessi ng	8	Medium	DWARAKNATH K, SWATHI K, HARINI K, DINESH KUMAR S, RAKSHANA S
Sprint-2	Model Building	USN-3	Import the required libraries, add the necessary layers and compile the model	10	High	DWARAKNATH K, SWATHI K, HARINI K, DINESH KUMAR S, RAKSHANA S
Sprint-2		USN-4	Training the image classific ation	7	Medium	DWARAKNATH K, SWATHI K, HARINI K, DINESH KUMAR S, RAKSHANA S

			model using CNN			
Sprint-3	Training and Testing	USN-5	Training the model and testing the model's performan ce	9	High	DWARAKNA TH K, SWATHI K, HARINI K, DINESH KUMAR S, RAKSHANA S
Sprint-4	Implementa tion of the application	USN-6	Conv erting the input sign langu age imag es into Engli sh alpha bets	8	Medium	DWARAKNA TH K, SWATHI K

6.2 Sprint Delivery Schedule

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

Sprint	Total Story Points	Durati on	Sprint Start Date	Sprint End Date (Planned)	Story Points Complet ed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-	10	6 Days	24 Oct 2022	29 Oct 2022	8	29 Oct 2022
Sprint-	10	6 Days	31 Oct 2022	04 Nov 2022	5	04 Nov 2022
Sprint-	10	6 Days	07 Nov 2022	11 Nov 2022	7	11 Nov 2022
Sprint-	10	6 Days	14 Nov 2022	18 Nov 2022	5	18 Nov 2022

Velocity:

Burndown chart:



SPRINT BURNDOWN CHART:

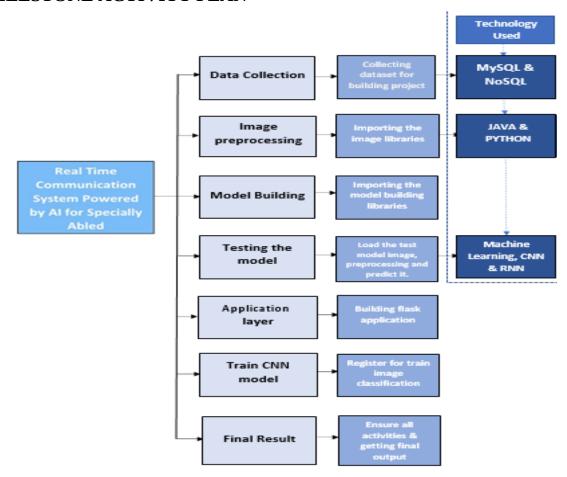


Milestone Activity Plan:

Milestone	Function (Epic)	Milesto	Story / Task	
		ne		
		Story		
		Numb		
		er		
			In order to build our	
Milestone 1	Data collection	M1	project, we are gathering	
			data and creating two	
			folders: onefor training and	
			the other for testing.	
			Importing libraries for	
Milestone2	Image	M2	image data generators and	
	preprocessi		using theirfeatures to	
	ng		trainthe test set.	
			Importing the libraries for	
Milestone3	Model building	M3	model construction, setting	
			up the model, Convolution	
			layeraddition, layer pooling,	
			layer flattening, layer dense	
			layer addition Putting	
			together the model and	
			saving it.	
			Initially import the	
Milestone4	Testing the	M4	packages. The test imageis	
	model		then preprocessed and	
			predicted whenthe model	
			has been	
			saved and loaded.	

Milestone 5	Application layer	M5	Develop the HTML pages and the Flaskapplication.
Milestone6	Train CNNmodel	M6	Activate IBM Cloud and prepare theimage classification model.
Milestone7	Final result	M7	To ensure that all processes result in thedesired outcome.

MILESTONE ACTIVITY PLAN



6.3 Reports from JIRA



7 CODING & SOLUTIONING

(Explain the features added in the project along with code)

7.1 Libraries to be installed

pip install fer
pip install flask
pip install cv2
pip install numpy
pip install keras
pip install tensorflow
pip install cvzone
pip install pyttsx3
pip install scikit-image

7.2 Real-Time Sign to Speech Translation

People who are unable to talk typically use sign language to communicate. The majority of people find it extremely challenging to communicate with silent persons since they are unable to grasp Universal Sign Language (unless they have studied it). The core of this project is a tool that allows silent people and others communicate with one another. Our technology makes use of a CNN-built model that can recognise sign languages in real time.

7.3 Real-Time Speech to Sign Conversion

We can use JavaScript to recognise voice thanks to the Web Speech API. JavaScript makes it very simple to recognise speech in a browser and then extract the text from the speech for use as user input. The voice is translated into text using the SpeechRecognition object, which is subsequently shown on the screen. This can be accomplished by our technology in real time. Any language that the user is trying to speak in can be recognised by it. But only the Chrome browser is supported for this API. The live example below may not function if you are viewing this example in another browser.

8 TESTING

8.1 Test Cases:

- Check to see that the options are visible when the user hits the URL.
- Check to see if the UI elements are displaying correctly.
- Check to see if the user can select any languages.
- Check to see if the user is being forwarded to the sign-to-speech website.
- Check to see if the program can convert the sign to speech.
- Check to make sure the user can leave the sign-to-speech page.

8.2 UAT Testing

Test Case Analysis

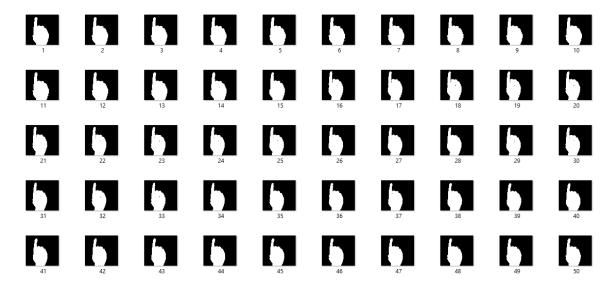
Section	Total Cases	Not Tested	Fail	Pass
Print Engine	7	0	0	7
Client Application	15	0	0	15
Security	2	0	0	2
Outsource Shipping	2	0	0	2
Exception Reporting	9	0	0	9
Final Report Output	4	0	0	4
Version Control	2	0	0	2

8.3 Performance Testing

S.NO	Parameter	Values	Screenshot		
1.	Model Summary				
		Total params: 1,103,721	Model: "sequential"		
		Trainable params:	Layer (type)		
		1,103,721	conv2d (Conv20)	(None, 62, 62, 32)	328
		Non-trainable params: 0	max_pooling2d (MaxPooling2D	(None, 31, 31, 32)	8
			conv2d_1 (Conv20)	(None, 29, 29, 512)	147968
			max_pooling2d_1 (MaxPooling 2D)	(None, 14, 14, 512)	0
			conv2d_2 (Conv20)	(None, 14, 14, 32)	147488
			max_pooling2d_2 (MaxPooling 2D)	(None, 7, 7, 32)	0
			flatten (Flatten)	(None, 1568)	0
			dense (Dense)	(None, 512)	803328
			dense_1 (Dense)	(None, 9)	4617
2.	Accuracy	Training Accuracy - 0.9994	,		
			Epoch 13/25 53/53 [] - 196s 4s/st	tep - loss: 0.0050 - accuracy: 0.9902 - val_loss:
		Validation Accuracy -0.9969	8.0647 - wal_accuracy: 0. Epoch 14/25 53/53 [====================================] - 196s 4s/st	tep - loss: 0.0024 - accuracy: 0.9990 - val_loss
			53/53 [====================================	9720	tep - loss: 0.8052 - accuracy: 0.9982 - val_loss:
			8.8172 - sal_sccurscy: 0. Epoch 17/25 53/53 [9969] - 195s 4s/st	tep - loss: 0.8422 - accuracy: 0.9994 - wal_loss tep - loss: 6.8413e-04 - accuracy: 0.9998 - wal_
			s: 0.1009 - val_accuracy:] - 196s 4s/st	tep - loss: 2.1259e-04 - accuracy: 1.0000 - val_
			Epoch 19/25 53/53 [tep - loss: 0.0036 - accuracy: 0.9987 - val_loss
			53/53 [9787	tep - loss: 0.8658 - accuracy: 0.9961 - val_loss tep - loss: 0.8634 - accuracy: 0.9967 - val_loss
			20/29	2 200 48/30	one comment and the state of th

```
from tensorflow.keras.models import load_model
from tensorflow.keras.preprocessing import image
import numpy as np
import cv2
                                                                     In []:
model = load model('/content/Real time.h5')
                                                                     In []:
img = image.load_img('/content/Dataset/test_set/H/107.png',target_size =
(100, 100))
img
                                                                    Out[]:
                                                                     In []:
from skimage.transform import resize
def detect(frame):
   img=image.img_to_array(frame)
   img = resize(img, (64, 64, 1))
   img = np.expand_dims(img, axis=0)
   pred=np.argmax(model.predict(img))
   op=['A','B','C','D','E','F','G','H','I']
   print("THE PREDICTED LETTER IS ", op[pred])
img=image.load_img("/content/Dataset/test_set/H/107.png")
detect(img)
1/1 [======] - 0s 28ms/step
THE PREDICTED LETTER IS H
                                                                     In []:
img = image.load_img('/content/Dataset/test_set/A/110.png')
pred=detect(img)
1/1 [======] - 0s 26ms/step
THE PREDICTED LETTER IS A
```

TEST SET: (ALPHABET D)



TRAINING SET (ALPHABET D)



9 RESULTS

Performance Metrics

Table 1. Performance Analysis

Training Data Set	Reference Image	Correlation Weight	Output
Imageone	Imageone	0.2458	Correctly Classified
imagetwo	imagetwo	0.2234	Correctly Classified
imageone	Imagetwo	1.2271	Incorrectly Classified
Imagetwo	imageone	1.0011	Incorrectly Classified



10 ADVANTAGES & DISADVANTAGES

ADVANTAGES

- 1. Sign language translation and detection in real time.
- 2. Developed model provides better accuracy.
- 3. Language is customizable.
- 4. User Interface is friendly.
- 5. There is real time speech to text conversion.
- 6. Data is secured and private.

DISADVANTAGES

- 1. The website would take some time to load.
- 2. Model still needs to be tested in a greater and macro level by wider range of data sets.
- 3. Customization of sign language is not available.
- 4. It is difficult to take notes while using the application.
- 5. The developed product works only in google chrome.
- 6. It is difficult to make calls using the developed application.

11 Conclusion:

With our method the frames are extracted from video. From the frames the hand gestures could be identified by using background subtraction techniques. The segmentation of the hand movements plays a vital role in such process. Sign languages, as spoken languages, have certain rules of grammar. These rules must be taken into account while transforming a sign language into a spoken language. In the end, adding a speech engine to speak the transformed text would help enhance ease of use. We are conducting further to develop a method for handling a large vocabulary. The detection capability of the system could be expanded to body gestures as well. The same technique could be used in a portable device.

12 Future Scope:

Researchers are working to improve hand gesture detection utilising the human nervous system. To recognise human gestures for automation, a system with a more sophisticated algorithm than those already presented and in use must be created.

13 Appendix

Source Code Model Building

import cv2
import os os.environ['TF_CPP_MIN_LOG_LEVEL'] = '2'
import numpy as np
from keras.models import Sequential

import matplotlib.pyplot as plt from keras.layers import Dense, Dropout,

```
Activation, Flatten
from keras.layers import Conv2D, MaxPool2D
from keras preprocessing.image import ImageDataGenerator test path =
'Dataset/test_set' train_path = 'Dataset/training_set'
train=ImageDataGenerator(rescale=1./255,zoom_range=0.2,shear_range=0.2,horiz
ontal flip=T rue) test=ImageDataGenerator(rescale=1./255)
train_batches = train.flow_from_directory(directory=train_path,
target_size=(64,64), class_mode='categorical',
batch size=300,shuffle=True,color mode="grayscale")
test_batches = test.flow_from_directory(directory=test_path, target_size=(64,64),
class_mode='categorical', batch_size=300, shuffle=True,color_mode="grayscale")
model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3), activation='relu',
input_shape=(64,64,1)))
model.add(MaxPool2D(pool size=(2,2)))
model.add(Conv2D(512, (3, 3), padding="valid"))
model.add(MaxPool2D(pool_size=(2,2)))
model.add(Conv2D(32, (3, 3), padding="same"))
model.add(MaxPool2D(pool_size=(2,2)))
model.add(Flatten())
model.add(Dense(512,activation ="relu"))
model.add(Dense(9,activation = "softmax"))
model.compile(optimizer='adam', loss='categorical_crossentropy',
metrics=['accuracy'])
history = model.fit(train batches,
batch_size=32,validation_data=test_batches,epochs=25) model.save('model.h5')
```

Model Testing

```
import keras from keras.models
import load_model
import cv2
import numpy as np
import os os.environ['TF_CPP_MIN_LOG_LEVEL'] = '2'
val=['A','B','C','D','E','F','G','H','I']
model=load_model('model.h5')
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
            predict_x=model.predict(img)
            print(predict_x)
            predict=np.argmax(predict_x,axis=1)
            x=predict[0] print(val[x])
frame=cv2.imread(r"C:\Users\Akshaya\PycharmProjects\Realtime_Communicati
on_System_For_Specially_Abled\Dataset\test_set\B\1.png")
data=detect(frame)
Camera.py
import cv2
import numpy as np
from tensorflow.keras.models import load_model
from tensorflow.keras.preprocessing import image
class Video(object):
```

```
def __init__(self):
            self.video = cv2.VideoCapture(0)
            self.roi_start = (50, 150)
            self.roi\_end = (250, 350)
            self.model = load_model('asl_model.h5') # Execute Local Trained
Model
            # self.model = load_model('IBM_Communication_Model.h5') #
Execute IBM Trained Model
            self.index=['A','B','C','D','E','F','G','H','I']
            self.y = None
      def __del__(self):
            self.video.release()
      def get_frame(self):
            ret,frame = self.video.read()
            frame = cv2.resize(frame, (640, 480))
            copy = frame.copy()
            copy = copy[150:150+200,50:50+200]
            # Prediction Start
Main.py
import cv2
video = cv2.VideoCapture(0)
while True:
      ret, frame = video.read()
      cv2.imshow("Frame", frame)
      k = cv2.waitKey(1)
      if k == ord('q'):
            break
video.release()
```

HTML File:

```
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="utf-8">
  <meta name="viewport" content="width=device-width, initial-scale=1.0, shrink-</pre>
to-fit=no">
  <title>SmartBridge_WebApp_VideoTemplate</title>
  <link rel="stylesheet"</pre>
href="https://cdn.jsdelivr.net/npm/bootstrap@5.1.3/dist/css/bootstrap.min.css">
  <link rel="stylesheet"</pre>
href="https://use.fontawesome.com/releases/v5.12.0/css/all.css">
  <link rel="stylesheet" href="assets/css/Banner-Heading-Image.css">
  <link rel="stylesheet" href="assets/css/Navbar-Centered-Brand.css">
  <link rel="stylesheet" href="assets/css/styles.css">
</head>
<body style="background: rgb(39,43,48);">
  <nav class="navbar navbar-light navbar-expand-md py-3" style="background:
#212529;">
     <div class="container">
       <div></div><a class="navbar-brand d-flex align-items-center"
href="#"><span
            class="bs-icon-sm bs-icon-rounded bs-icon-primary d-flex justify-
content-center align-items-center me-2 bs-icon"><i
              class="fas fa-flask"></i></span><span style="color:
rgb(255,255,255);">Real-Time Communication
            System Powered By AI For Specially Abled</span></a>
```

```
<div></div>
     </div>
  </nav>
  <section>
    <div class="d-flex flex-column justify-content-center align-items-center">
       <div class="d-flex flex-column justify-content-center align-items-center"</pre>
id="div-video-feed"
         style="width: 640px;height: 480px;margin: 10px;min-height: 480px in-
width: 640px;border-radius: 10px;border: 4px dashed rgb(255,255,255);">
         <img src="{{ url_for('video_feed') }}" style="width: 100%;height:</pre>
100%;color: rgb(255,255,255);text-align: center;font-size: 20px;"
            alt="Camera Access Not Provided!">
       </div>
     </div>
     <div class="d-flex flex-column justify-content-center align-items-center"</pre>
style="margin-bottom: 10px;"><button
         class="btn btn-info" type="button" data-bs-target="#modal-1" data-bs-
toggle="modal">Quick Reference
         -<strong> ASL Alphabets</strong></button></div>
  </section>
  <section>
     <div class="container">
       <div class="accordion text-white" role="tablist" id="accordion-1">
          <div class="accordion-item" style="background: rgb(33,37,41);">
            <h2 class="accordion-header" role="tab"><button class="accordion-
button" data-bs-toggle="collapse"
                 data-bs-target="#accordion-1 .item-1" aria-expanded="true"
                 aria-controls="accordion-1 .item-1"
                 style="background: rgb(39,43,48);color:
rgb(255,255,255);">About The Project</button></h2>
```

```
<div class="accordion-collapse collapse show item-1" role="tabpanel"</pre>
data-bs-parent="#accordion-1">
              <div class="accordion-body">
                 Artificial Intelligence has made it possible to
handle our daily activities
                   in new and simpler ways. With the ability to automate tasks
that normally require human
                   intelligence, such as speech and voice recognition, visual
perception, predictive text
                   functionality, decision-making, and a variety of other tasks, AI
can assist people with
                   disabilities by significantly improving their ability to get
around and participate in
                   daily activities. <br > Currently, Sign Recognition is
available <strong>only for
                      alphabets A-I</strong> and not for J-Z, since J-Z alphabets
also require Gesture
                   Recognition for them to be able to be predicted correctly to a
certain degree of
                   accuracy.
              </div>
            </div>
          </div>
          <div class="accordion-item" style="background: rgb(33,37,41);">
            <h2 class="accordion-header" role="tab"><button class="accordion-
button collapsed"
                 data-bs-toggle="collapse" data-bs-target="#accordion-1 .item-2"
aria-expanded="false"
                 aria-controls="accordion-1 .item-2"
                 style="background: rgb(39,43,48);color:
```

```
rgb(231,241,255);">Developed By</button></h2>
           <div class="accordion-collapse collapse item-2" role="tabpanel" data-</pre>
bs-parent="#accordion-1">
              <div class="accordion-body">
                Students at VIT-Bhopal University during
SmartBridge AI Externship
                  Program. <br/>
<br/>br>1. <strong>Nirlov Deb</strong>
19BCG10067<br>2.
                  <strong>Kushagra</strong> 19BCG10025<br>>3.
<strong>Kartik Dhasmana</strong> 19BCG10002
                </div>
           </div>
         </div>
       </div>
    </div>
  </section>
  <div class="modal fade" role="dialog" tabindex="-1" id="modal-1">
    <div class="modal-dialog" role="document">
       <div class="modal-content">
         <div class="modal-header">
           <h4 class="modal-title">American Sign Language -
Alphabets</h4><button type="button"
              class="btn-close" data-bs-dismiss="modal" aria-
label="Close"></button>
         </div>
         <div class="modal-body"><img src="{{ url_for('static',</pre>
filename='img/ASL_Alphabets.png') }}" width="100%"></div>
         <div class="modal-footer"><button class="btn btn-secondary"</pre>
type="button"
```

```
data-bs-dismiss="modal">Close</button></div>
       </div>
     </div>
  </div>
  <script
src="https://cdn.jsdelivr.net/npm/bootstrap@5.1.3/dist/js/bootstrap.bundle.min.js">
</script>
</body>
</html>
cv2.destroyAllWindows()
            cv2.imwrite('image.jpg',copy)
            copy_img = image.load_img('image.jpg', target_size=(64,64))
            x = image.img_to_array(copy_img)
            x = np.expand_dims(x, axis=0)
            pred = np.argmax(self.model.predict(x), axis=1)
            self.y = pred[0]
            cv2.putText(frame, The Predicted Alphabet is:
'+str(self.index[self.y]),(100,50),cv2.FONT_HERSHEY_SIMPLEX,1,(0,0,0),3)
            ret,jpg = cv2.imencode('.jpg', frame)
            return jpg.tobytes()
```

CSS File:

Navbar Centered Brand.Css

```
.bs-icon {
--bs-icon-size: .75rem;
display: flex;
flex-shrink: 0;
justify-content: center;
align-items: center;
font-size: var(--bs-icon-size);
width: calc(var(--bs-icon-size) * 2);
height: calc(var(--bs-icon-size) * 2);
color: var(--bs-primary);
}
.bs-icon-xs {
 --bs-icon-size: 1rem;
 width: calc(var(--bs-icon-size) * 1.5);
 height: calc(var(--bs-icon-size) * 1.5);
}
.bs-icon-sm {
 --bs-icon-size: 1rem;
}
.bs-icon-md {
 --bs-icon-size: 1.5rem;
}
.bs-icon-lg {
```

```
--bs-icon-size: 2rem;
.bs-icon-xl {
 --bs-icon-size: 2.5rem;
}
.bs-icon.bs-icon-primary {
 color: var(--bs-white);
 background: var(--bs-primary);
}
.bs-icon.bs-icon-primary-light {
 color: var(--bs-primary);
 background: rgba(var(--bs-primary-rgb), .2);
}
.bs-icon.bs-icon-semi-white {
 color: var(--bs-primary);
 background: rgba(255, 255, 255, .5);
}
.bs-icon.bs-icon-rounded {
 border-radius: .5rem;
}
.bs-icon.bs-icon-circle {
 border-radius: 50%;
}
```

Output 13.2

Github and Demo Link:

https://github.com/IBM-EPBL/IBM-Project-17195-1659630234