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1. INTRODUCTION

1.1 Project Overview

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. Real-time communications (RTC) is any mode of telecommunications in which all users can exchange information instantly.

Communication plays a significant role in making the world better place. It creates a bonding and relations among the people.

1.2 Purpose

The project aims to develop a system that converts the sign language into a human hearing voice in the desired language to convey a message to normal people, as well as convert speech into understandable sign language for the deaf and dumb using the convolutional neural network.

An app is built which enables the deaf and dumb people to convey their information using signs which is converted to human understandable language and output is given as speech.

2. LITERATURE SURVEY

2.1 Existing problem

Communications between deaf-mute and a normal person has always been a challenging task. It is very difficult for mute people to convey their message to normal people. Since normal people are not trained on hand sign language.

Only specially abled people are taught sign language and the common person is unaware its working causing a communication gap. Under emergency situations, it is even more difficult for specially abled people to get help. Non-Emergency normal environments can also be hard for them to navigate needing special assistance.

2.2 References

1. Upendran, S., and Thamizharasi, A., "American Sign Language interpreter system for deaf and dumb individuals", In the Proceedings of the International Conference on Control, Instrumentation, Communication and Computational Technologies (ICCICCT), pp. 1477-1481, 2014
2. Lotti, F., Tiezzi, P., Vassura, G., Biagiotti, L., and Melchiorri, C., "UBH 3: an anthropomorphic hand with simplified endo-skeletal structure and soft continuous fingerpads", In Proceedings IEEE International Conference on Robotics and Automation, 2004 (ICRA'04), Vol.5, pp. 4736-474, IEEE, 2004.
3. Rajamohan, A., Hemavathy, R., and Dhanalakshmi, M., "Deaf-Mute Communication Interpreter", International Journal of Scientific Engineering and Technology, Vol.2, No.5, pp.336-341, 2013.

4. Verma, P., Shimi S. L. and Priyadarshani, R., "Design of Communication Interpreter for Deaf and Dumb Person", Vol.4, no.1, 2013.

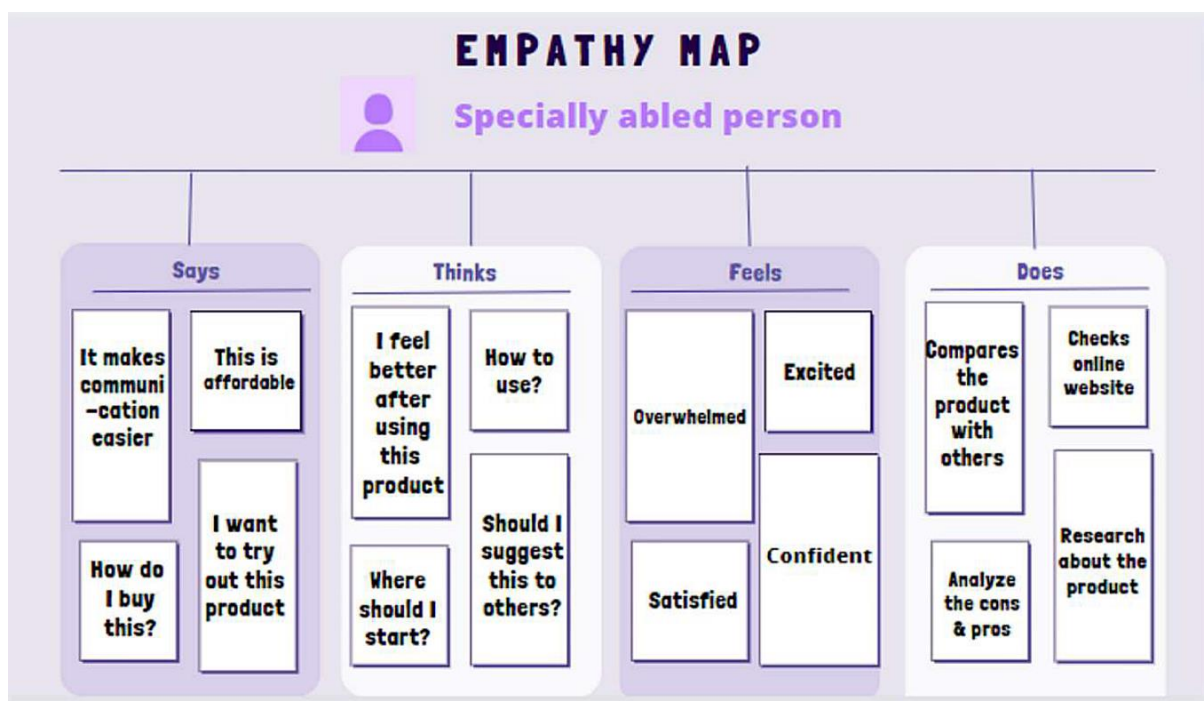
2.3 Problem Statement Definition

Only specially abled people are taught sign language and the common person is unaware its working causing a communication gap. Under emergency situations, it is even more difficult for specially abled people to get help. Non-Emergency normal environments can also be hard for them to navigate needing special assistance.

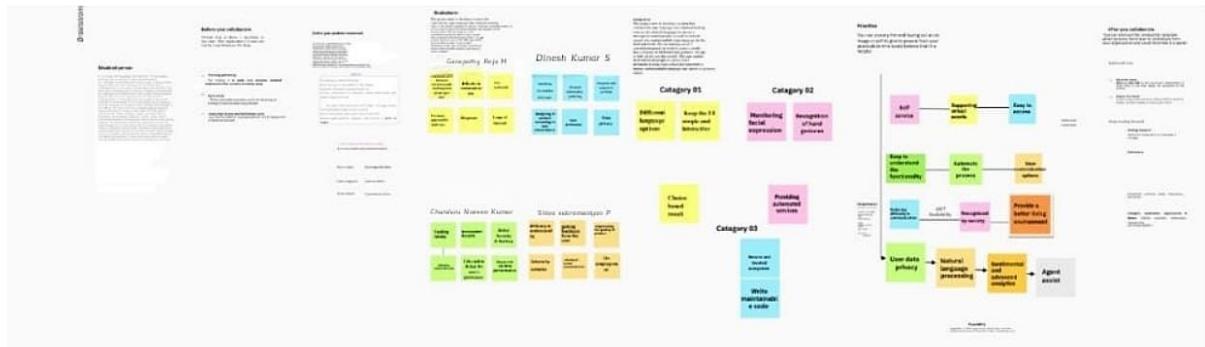
Communications between deaf-mute and a normal person has always been a challenging task. It is very difficult for mute people to convey their message to normal people. Since normal people are not trained on hand sign language.

3. IDEATION & PROPOSED SOLUTION

3.1 Empathy Map Canvas



3.2 Ideation & Brainstorming



3.3 Proposed Solution

SNO	Parameter	Description
1.	Problem Statement(Problem to be solved)	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 very difficult 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 very useful to have a proper conversation between a normal person and an impaired person in any language
2.	Idea / Solution description	The project aims to develop a system that converts the sign language into a human hearing voice in the desired language to convey a message to normal people, as well as convert speech into understandable sign language for the deaf and dumb. We are making use of a convolution neural network to create a model that is trained on different hand gestures. An app is built which uses this model. This app enables deaf and dumb people to convey their information using signs which get converted to human-understandable language and speech is given as output.
3.	Novelty / Uniqueness	Building mobile tools with data isn't as easy as importing an XML feed of your latest headlines. But if you're going to spend thousands of dollars developing a mobile app anyway, you might as well spend a little more to build a real application that helps solve problems and makes advertisers take notice.

4.	Social Impact / Customer Satisfaction	These apps are using only for solve the problems. There are many different types of disabilities, and there are also many different ways in which people may use AI. The use of artificial intelligence is a boon for specially-abled people. Technology had opened up new opportunities and created jobs where none had existed before, such as speech to text software that helped one woman find her voice after she was paralysed in an accident. Artificial Intelligence can help those with disabilities accomplish tasks they never thought possible; here are just a few ways we've seen AI technology impact lives: Facial recognition and predictive texting tools
5.	Business Model (Revenue Model)	Building mobile tools with data isn't as easy as importing an XML feed of your latest headlines. But if you're going to spend thousands of dollars developing a mobile app anyway, you might as well spend a little more to build a real application that helps solve problems and makes advertisers take notice
6.	Scalability of the Solution	With the help of machine tasks that usually requires human intelligence, such as voice and speech synthesis, visual perception, predictive text functionality, judgement, and a variety of other tasks, AI can assist individuals with disabilities by making a significant distinction in their ability.

3.4 Problem Solution

1. CUSTOMER SEGMENT(S) Specially abled person	6. CUSTOMER CONSTRAINTS spending power, budget, no cash, network connection, available devices.	5. AVAILABLE SOLUTIONS In this application, we provide feedback pop-ups frequently and an emergency ping for people who have minimum knowledge about the application.
2. JOBS-TO-BE-DONE / PROBLEMS Concentrate on making their communication much easier and live a normal life.	9. PROBLEM ROOT CAUSE Mostly genetic problems and some problems may be caused because of some viral infections. Poverty and malnutrition, accidents, poor health care.	7. BEHAVIOUR The customer will be provided with a customer care number and also there will be many feedback pop-ups frequently which helps the customer to contact us and get their jobs done.
3. TRIGGERS Advertising the product in specially abled schools and create awareness about this product the help of government. And also advertise using various online platforms such as you tube, face book etc. 4. EMOTIONS: BEFORE / AFTER Before: Before using this product specially abled people struggled to communicate with others. After: But after using this product they will feel easy and comfortable to communicate. They feel more confident about themselves.	10. YOUR SOLUTION Facial recognition, Voice recognition and predictive texting tools allows people who have difficulties in speaking to communicate more easily using Artificial Intelligence. We can also use AI Sensors to monitor their health conditions regularly and save the health reports for future purposes in a separate database.	8. CHANNELS of BEHAVIOUR 8.1 ONLINE So, in online the customer will use various online voice assistant technologies such as Siri, Alexa, Google Assistants etc. 8.2 OFFLINE Connecting with people might be difficult depending on the type of disabilities. So, technology and AI leaves no one behind and can benefit persons with impairments.

4. REQUIREMENT ANALYSIS

4.1 Functional Requirement

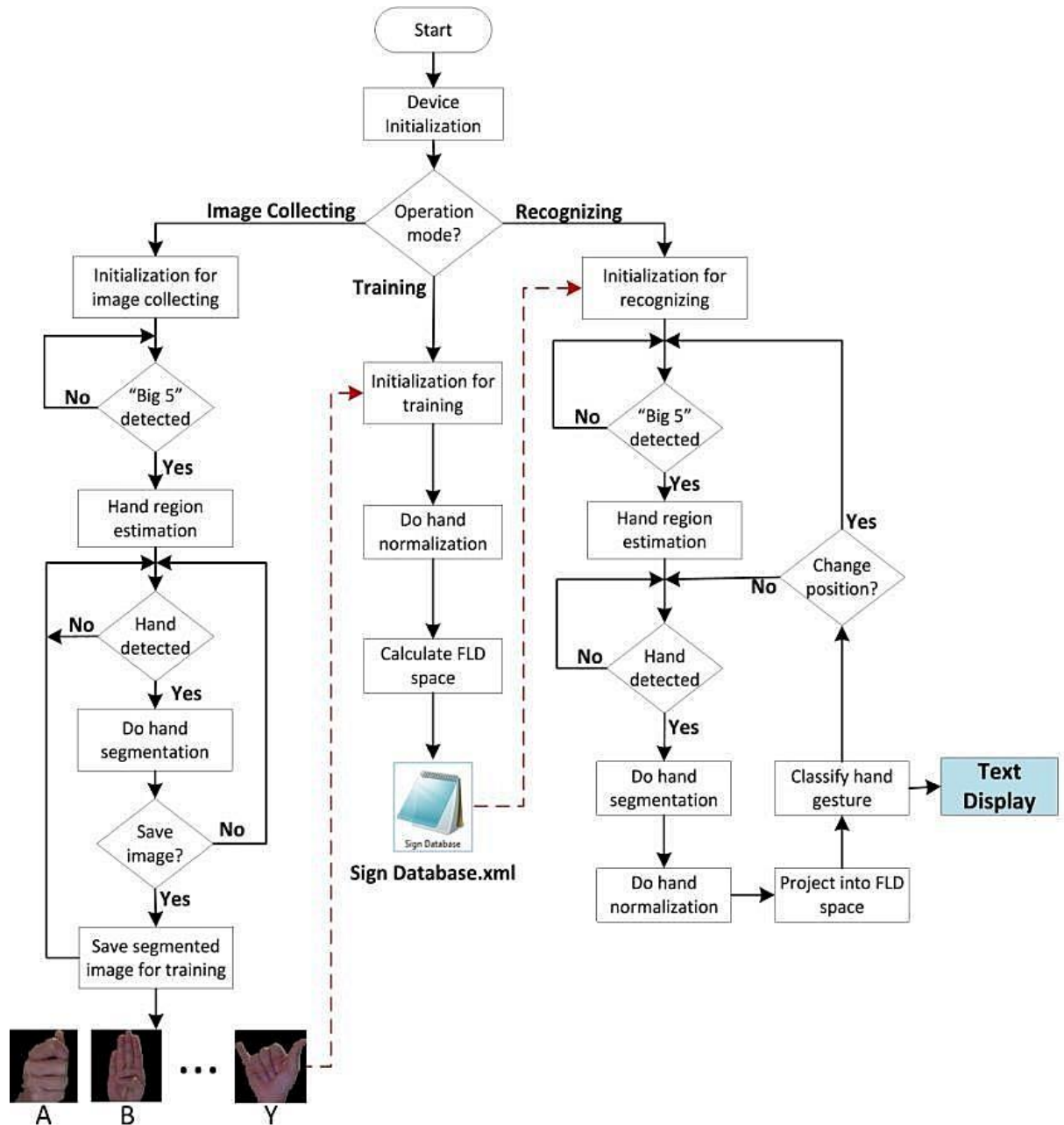
FR NO.	Functional Requirement(Epic)	Sub Requirement(Story/Sub-Task)
FR-1	User Registration	Registration through From Registration through Gmail Registration through Linked IN
FR-2	User Confirmation	Confirmation Via Email Confirmation Via OTP
FR-3	Uploading Image	Upload image through camera or Upload image through gallery
FR-4	Templates usage During any Emergencies	Select emergency templates icon to pass the message quickly
FR-5	Text to Speech	Convert Respective text to sign Language

4.2 Non Functional Requirement

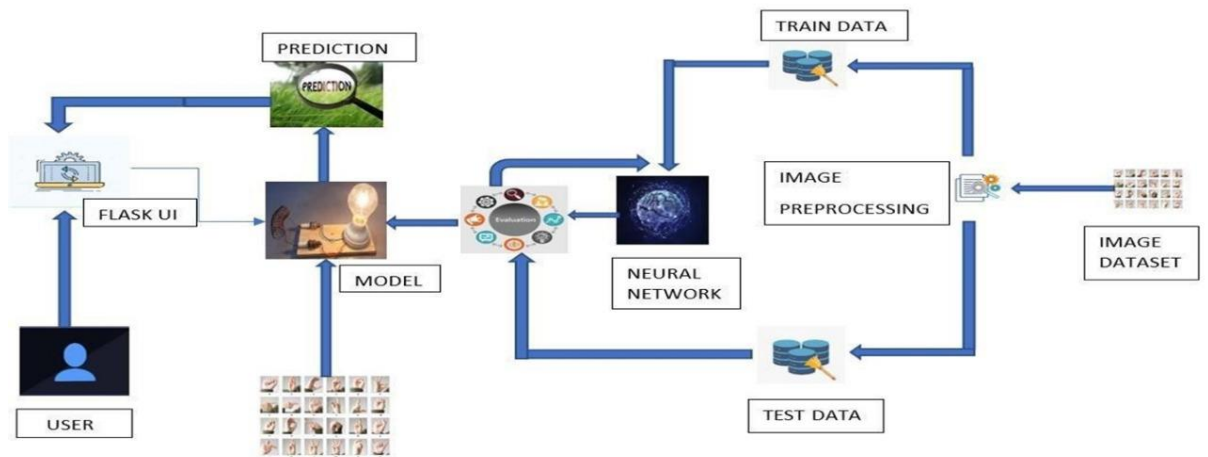
FR NO.	Non- Functional Requirement	Description
NFR-1	Usability	user can easily upload the image and this app is designed in a way where user can easily find some predefined templates or layouts.
NFR-2	Security	user should sign in into the application. so any unauthorized access will be avoided at the most.
NFR-3	Reliability	This application has robust fault tolerance and even if there is an error, it recuperates swiftly.
NFR-4	Performance	Utilizing the CNN model, the gestures made by the user is predicted by the application with a higher accuracy.
NFR-5	Availability	This is application is effortless and is accessible to all users. The predefined formats or layouts are understandable and makes it easier for accessing it.
NFR-6	Scalability	Highly scalable which uses gesture recognition and is a hands on model which is used in many other applications.

5. PROJECT DESIGN

5.1 Data Flow Diagrams



5.2 Solution & Technical Architecture



5.3 User Stories

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer (Mobile user)	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming the password	I can access my account / dashboard	High	Sprint-1
		USN-2	As a user, I will receive confirmation email once I have registered for the application	I can receive confirmation email & click confirm	High	Sprint-1
		USN-3	As a user, I can register for the application through Facebook	I can register & access the dashboard with Facebook Login	Low	Sprint-2
		USN-4	As a user, I can register for the application through Gmail	I can register via some third party's link	Medium	Sprint-1
	Login	USN-5	As a user, I can log into the application by entering email & password	I can type manually and also can use saved login credentials	High	Sprint-1
	Dashboard	USN-6	As a customer, I can get all services and help through the dashboard	I can access my dashboard and change profile	Medium	Sprint-2
Customer (Web user)	Registration	USN-7	As a customer, I could be able to login through registered phone number by using OTP instead of Gmail	I could be able to register & login via phone number to access my account	High	Sprint-2
Customer Care Executive	Service	USN-8	Can avail the service by calling customer care or reaching through E-mail.	Can avail the service by calling customer care or reaching through E-mail.	Medium	Sprint-1
Administrator	Sign up	USN-9	Customer have to sign-up to use these things and all	Have to enter valid credentials.	High	Sprint-1
	Enrollment	USN-10	The customer can avail all services once he/she enrolled.	As customer it is quite enchanting.	Medium	Sprint-2

6. PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning & Estimation

Milestone	Function (Epic)	Milestone Story Number	Story / Task
Milestone 1	Data collection	M1	Collected dataset for building our project and created two folders: testing and training.
Milestone 2	Image preprocessing	M2	Importing image data generator libraries and applying image data generator functionality to train the test set.
Milestone 3	Model building	M3	Importing the model building libraries, Initializing the model, Adding Convolution layers, Adding the Pooling layers, Adding the Flatten layers, Adding Dense layers, Compiling the model Fit and Save the model.
Milestone 4	Testing the model	M4	Import the packages first. Save the model and Load the test image, preprocess it and predict it.
Milestone 5	Application layer	M5	Build the flask application and HTML pages.
Milestone 6	Train CNN model	M6	Register for IBM Cloud and train Image Classification Model.
Milestone 7	Final result	M7	Final output was verified.

6.2 Sprint Delivery Schedule

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Registration	USN-1	As a user, I can register for the application by entering my email, password, and confirming my password.	5	High	GANAPATHY RAJA CHUNDURU NAVEEN KUMAR SHIVA SUBRAMANIYAN DINESH KUMAR
	Login	USN-2	As a user, I can log into the application by entering email & password	5	High	GANAPATHY RAJA CHUNDURU NAVEEN KUMAR SHIVA SUBRAMANIYAN DINESH KUMAR
	Data Collection	USN-3	Collecting Dataset	5	High	GANAPATHY RAJA CHUNDURU NAVEEN KUMAR SHIVA SUBRAMANIYAN DINESH KUMAR
	Image pre-processing	USN-4	Perform pre-processing techniques on the dataset	5	High	GANAPATHY RAJA CHUNDURU NAVEEN KUMAR SHIVA SUBRAMANIYAN DINESH KUMAR
Sprint-2	Model Building	USN-5	Model initialization with required layers	5	High	GANAPATHY RAJA CHUNDURU NAVEEN KUMAR SHIVA SUBRAMANIYAN DINESH KUMAR
	Training	USN-6	Training the image classification model using CNN	5	High	GANAPATHY RAJA CHUNDURU NAVEEN KUMAR SHIVA SUBRAMANIYAN DINESH KUMAR
Sprint-3	Testing	USN-7	Testing the model's performance	5	High	GANAPATHY RAJA CHUNDURU NAVEEN KUMAR SHIVA SUBRAMANIYAN DINESH KUMAR
Sprint-4	Deployment of model in web/app	USN-8	Converting text to speech API	5	High	GANAPATHY RAJA CHUNDURU NAVEEN KUMAR SHIVA SUBRAMANIYAN DINESH KUMAR

7. CODING & SOLUTIONING (Explain the features added in the project along with code)

7.1 Model Building

Importing The Required Model Building Libraries

```
In [1]: from tensorflow.keras.preprocessing.image import ImageDataGenerator

In [ ]: # Training Datagen
train_datagen = ImageDataGenerator(rescale=1/255, zoom_range=0.2, horizontal_flip=True, vertical_flip=False)
# Testing Datagen
test_datagen = ImageDataGenerator(rescale=1/255)

In [ ]: #training Dataset
x_train=train_datagen.flow_from_directory('/content/training_set', target_size=(64,64), batch_size=200,
                                         class_mode='categorical', color_mode="grayscale")
x_test=test_datagen.flow_from_directory('/content/test_set', target_size=(64,64), batch_size=200,
                                       class_mode='categorical', color_mode="grayscale")

Found 15130 images belonging to 9 classes.
Found 2250 images belonging to 9 classes.

In [ ]: import tensorflow as tf
import os
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Conv2D, Flatten, Dropout, MaxPooling2D
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import numpy as np
import matplotlib.pyplot as plt
import IPython.display as display
from PIL import Image
import pathlib

In [ ]: !unzip /content/test_set.zip

Archive: /content/test_set.zip
  creating: test_set/A/
   inflating: test_set/A/1.png
   inflating: test_set/A/10.png
 extracting: test_set/A/100.png
   inflating: test_set/A/101.png
   inflating: test_set/A/102.png
 extracting: test_set/A/103.png
   inflating: test_set/A/104.png
   inflating: test_set/A/105.png
   inflating: test_set/A/106.png
   inflating: test_set/A/107.png
   inflating: test_set/A/108.png
   inflating: test_set/A/109.png
   inflating: test_set/A/11.png
   inflating: test_set/A/110.png
 inflating: test set/A/111.one
```

Applying ImageDataGenerator Functionality to Trainset and Testset

```
In [ ]: x_train=train_datagen.flow_from_directory('/content/training_set',target_size=(64,64),batch_size=300,class_mode='categorical',color_mode="grayscale")
```

Found 15130 images belonging to 9 classes.

```
In [ ]: x_test=test_datagen.flow_from_directory('/content/test_set',target_size=(64,64),batch_size=300,class_mode='categorical',color_mode="grayscale")
```

Found 2250 images belonging to 9 classes.

```
In [ ]: print("Len x-train : ",len(x_train))
        print("Len x-test : ", len(x_test))
```

Len x-train : 51
Len x-test : 8

```
In [ ]: # The Class Indices in Training Dataset
        x_train.class_indices
```

Out[]: {'A': 0, 'B': 1, 'C': 2, 'D': 3, 'E': 4, 'F': 5, 'G': 6, 'H': 7, 'I': 8}

```
In [ ]: datagen=ImageDataGenerator(rescale=1./255,validation_split=0.25)
```

Model Creation

```
In [ ]: # Importing Libraries

        from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Convolution2D,MaxPooling2D,Flatten,Dense
```

```
In [ ]: # Creating Model
        model=Sequential()
```

```
In [ ]: from keras.preprocessing.image import ImageDataGenerator
        train_datagen=ImageDataGenerator(rescale = 1./255, shear_range=0.2, zoom_range=0.2,horizontal_flip=True,vertical_flip=False)
        test_datagen = ImageDataGenerator(rescale=1./255)
```

```
In [ ]: x_train = train_datagen.flow_from_directory("/content/training_set", target_size=(64,64),batch_size=100,
        class_mode='categorical', color_mode = "grayscale")
```

Found 15130 images belonging to 9 classes.

```
In [ ]: x_test = test_datagen.flow_from_directory("/content/test_set", target_size=(64,64),batch_size=100,
                                             class_mode='categorical', color_mode="grayscale")
```

Found 2250 images belonging to 9 classes.

```
In [ ]: len(x_train)
```

Out[]: 152

```
In [ ]: len(x_test)
```

Out[]: 23

```
In [ ]: x_train.class_indices
```

Out[]: {'A': 0, 'B': 1, 'C': 2, 'D': 3, 'E': 4, 'F': 5, 'G': 6, 'H': 7, 'I': 8}

MODEL BUILDING

```
In [ ]: from keras.models import Sequential
from keras.layers import Dense
from keras.layers import Convolution2D
from tensorflow.keras.layers import Conv2D, MaxPooling2D
from keras.layers import Dropout
from keras.layers import Flatten
```

```
In [ ]: #Creating the model
model=Sequential()
#Adding the layers
model.add(Convolution2D(32,(3,3), input_shape=(64,64,1), activation = 'relu'))
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Flatten())

#adding hidden layers
model.add(Dense(400, activation='relu'))
model.add(Dense(200, activation='relu'))
model.add(Dense(100, activation='relu'))

#Adding the output layer
model.add(Dense(9, activation='softmax'))
```

```
In [ ]: model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
```

```
In [ ]: model.fit_generator(x_train, steps_per_epoch=30, epochs=10, validation_data=x_test,validation_steps=50)
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: UserWarning: `Model.fit_generator` is deprecated and will be removed in a future version. Please use `Model.fit`, which supports generators.
"""Entry point for launching an IPython kernel.
Epoch 1/10
30/30 [=====] - ETA: 0s - loss: 1.0847 - accuracy: 0.6197
WARNING:tensorflow:Your input ran out of data; interrupting training. Make sure that your dataset or generator can generate at least `steps_per_epoch * epochs` batches (in this case, 50 batches). You may need to use the repeat() function when building your dataset.
30/30 [=====] - 19s 570ms/step - loss: 1.0847 - accuracy: 0.6197 - val_loss: 0.4876 - val_accuracy: 0.8582
Epoch 2/10
30/30 [=====] - 13s 435ms/step - loss: 0.2593 - accuracy: 0.9218
Epoch 3/10
30/30 [=====] - 14s 468ms/step - loss: 0.1441 - accuracy: 0.9583
Epoch 4/10
30/30 [=====] - 13s 441ms/step - loss: 0.1015 - accuracy: 0.9673
Epoch 5/10
30/30 [=====] - 13s 442ms/step - loss: 0.0687 - accuracy: 0.9780
Epoch 6/10
30/30 [=====] - 15s 494ms/step - loss: 0.0438 - accuracy: 0.9863
Epoch 7/10
30/30 [=====] - 13s 446ms/step - loss: 0.0549 - accuracy: 0.9820
Epoch 8/10
30/30 [=====] - 13s 442ms/step - loss: 0.0412 - accuracy: 0.9887
Epoch 9/10
30/30 [=====] - 13s 446ms/step - loss: 0.0276 - accuracy: 0.9920
Epoch 10/10
30/30 [=====] - 15s 487ms/step - loss: 0.0165 - accuracy: 0.9959
```

```
Out[ ]:
```

Add The Convolution Layer

```
In [ ]: model.save('Real_time.h5')
```

```
In [ ]: a=len(x_train)
        b=len(x_test)
```

```
In [ ]: print(a)
```

```
152
```

```
In [ ]: print(b)
```

```
23
```

```
In [ ]: #create model
        model=Sequential()
```

```
In [ ]: model.add(Convolution2D(32,(3,3),input_shape=(64,64,1),activation='relu'))
```

Add the Pooling Layer

```
In [ ]: model.add(MaxPooling2D(pool_size=(2,2)))
```

Add the Flatter Layer

```
In [ ]: model.add(Flatten())
```

The_Dense_Layers

```
In [ ]: #1st hidden Layer
        model.add(Dense(units=512,activation='relu'))
        #2nd hidden Layer
        model.add(Dense(units=256,activation='relu'))
```

```
In [ ]: #output Layer
        model.add(Dense(units=9,activation='softmax'))
```

Compile the Model

```
In [ ]: model.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['accuracy'])
```

Fit and Save the Model

```
In [ ]: model.save('aslpng2.h5')
```

Loading the Dataset& Image Data Generation

```
In [ ]: from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
In [ ]: # Training Datasets
train_datagen = ImageDataGenerator(rescale=1/255, zoom_range=0.2, horizontal_flip=True, vertical_flip=False)
# Testing Datasets
test_datagen = ImageDataGenerator(rescale=1/255)
```

```
In [ ]: #training Dataset
x_train=train_datagen.flow_from_directory('/content/training_set',target_size=(64,64),batch_size=200,
                                         class_mode='categorical',color_mode="grayscale")
x_test=test_datagen.flow_from_directory('/content/test_set',target_size=(64,64),batch_size=200,
                                       class_mode='categorical',color_mode="grayscale")
```

Found 15130 images belonging to 9 classes.
Found 2250 images belonging to 9 classes.

```
In [ ]: print("len x-train :", len(x_train))
        print("len x-test :", len(x_test))
```

len x-train : 76
len x-test : 12

```
In [ ]: # The Class Indices in Training Dataset
x_train.class_indices
```

```
Out[ ]: {'A': 0, 'B': 1, 'C': 2, 'D': 3, 'E': 4, 'F': 5, 'G': 6, 'H': 7, 'I': 8}
```

Model Creation

```
In [ ]: #Importing Libraries
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Convolution2D,MaxPooling2D,Flatten,Dense
```

```
In [ ]: #create model
model=Sequential()
```

```
In [ ]: #Adding Layers
model.add(Convolution2D(32,(3,3),input_shape=(64,64,1),activation='relu'))
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Flatten())

#Adding hidden Layer
model.add(Dense(units=512,activation='relu'))
model.add(Dense(units=261,activation='relu'))

#output Layer
model.add(Dense(units=9,activation='softmax'))
```

```
In [ ]: #Compiling the Model
model.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['accuracy'])
```

```
In [ ]: #Fitting the model
model.fit_generator(x_train, steps_per_epoch=30, epochs=10, validation_data=x_test,validation_steps=50)
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:2: UserWarning: 'Model.fit_generator' is deprecated and will be removed in a future version. Please use 'Model.fit', which supports generators.

Epoch 1/10
30/30 [=====] - ETA: 0s - loss: 0.6843 - accuracy: 0.7643

WARNING:tensorflow:Your input ran out of data; interrupting training. Make sure that your dataset or generator can generate at least `steps_per_epoch * epochs` batches (in this case, 50 batches). You may need to use the repeat() function when building your dataset.

30/30 [=====] - 39s 1s/step - loss: 0.6843 - accuracy: 0.7643 - val_loss: 0.3408 - val_accuracy: 0.9071

Epoch 2/10
30/30 [=====] - 35s 1s/step - loss: 0.1345 - accuracy: 0.9608

Epoch 3/10
30/30 [=====] - 33s 1s/step - loss: 0.0763 - accuracy: 0.9798

Epoch 4/10
30/30 [=====] - 32s 1s/step - loss: 0.0469 - accuracy: 0.9862

Epoch 5/10
30/30 [=====] - 32s 1s/step - loss: 0.0304 - accuracy: 0.9922

Epoch 6/10
30/30 [=====] - 33s 1s/step - loss: 0.0225 - accuracy: 0.9935

Epoch 7/10
30/30 [=====] - 35s 1s/step - loss: 0.0117 - accuracy: 0.9968

Epoch 8/10
30/30 [=====] - 32s 1s/step - loss: 0.0108 - accuracy: 0.9967

Epoch 9/10
30/30 [=====] - 32s 1s/step - loss: 0.0168 - accuracy: 0.9947

Epoch 10/10
30/30 [=====] - 33s 1s/step - loss: 0.0178 - accuracy: 0.9952

8. TESTING

8.1 TEST CASES

Loading the Dataset& Image Data Generation

```
In [ ]: from tensorflow.keras.preprocessing.image import ImageDataGenerator

In [ ]: # Training Datagen
train_datagen = ImageDataGenerator(rescale=1/255, zoom_range=0.2, horizontal_flip=True, vertical_flip=False)
# Testing Datagen
test_datagen = ImageDataGenerator(rescale=1/255)

In [ ]: #training Dataset
x_train=train_datagen.flow_from_directory('/content/training_set',target_size=(64,64),batch_size=200,
                                         class_mode='categorical',color_mode="grayscale")
x_test=test_datagen.flow_from_directory('/content/test_set',target_size=(64,64),batch_size=200,
                                       class_mode='categorical',color_mode="grayscale")

Found 15130 images belonging to 9 classes.
Found 2250 images belonging to 9 classes.

In [ ]: print("len x-train :", len(x_train))
        print("len x-test :", len(x_test))

len x-train : 76
len x-test : 12

In [ ]: # The Class Intices in Training Dataset
x_train.class_indices
```

Out[]: {'A': 0, 'B': 1, 'C': 2, 'D': 3, 'E': 4, 'F': 5, 'G': 6, 'H': 7, 'I': 8}

Model Creation

```
In [ ]: #Importing Libraries
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Convolution2D,MaxPooling2D,Flatten,Dense

In [ ]: #create model
model=Sequential()

In [ ]: #Adding Layers
model.add(Convolution2D(32,(3,3),input_shape=(64,64,1),activation='relu'))
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Flatten())

#Adding hidden layer
model.add(Dense(units=512,activation='relu'))
model.add(Dense(units=261,activation='relu'))

#output layer
model.add(Dense(units=9,activation='softmax'))

In [ ]: #Compiling the Model
model.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['accuracy'])

In [ ]: #Fitting the model
model.fit_generator(x_train, steps_per_epoch=30, epochs=10, validation_data=x_test,validation_steps=50)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:2: UserWarning: `Model.fit_generator` is deprecated and will be removed in a future version. Please use `Model.fit`, which supports generators.

Epoch 1/10
30/30 [=====] - ETA: 0s - loss: 0.6843 - accuracy: 0.7643
WARNING:tensorflow:Your input ran out of data; interrupting training. Make sure that your dataset or generator can generate at least `steps_per_epoch * epochs` batches (in this case, 50 batches). You may need to use the repeat() function when building your dataset.
30/30 [=====] - 39s 1s/step - loss: 0.6843 - accuracy: 0.7643 - val_loss: 0.3408 - val_accuracy: 0.9071
Epoch 2/10
30/30 [=====] - 35s 1s/step - loss: 0.1345 - accuracy: 0.9608
Epoch 3/10
30/30 [=====] - 33s 1s/step - loss: 0.0763 - accuracy: 0.9798
Epoch 4/10
30/30 [=====] - 32s 1s/step - loss: 0.0469 - accuracy: 0.9862
Epoch 5/10
30/30 [=====] - 32s 1s/step - loss: 0.0304 - accuracy: 0.9922
Epoch 6/10
```

8.2 User Acceptance Testing

1. Purpose of Document

The purpose of this document is briefly explain the test coverage and open issues of the [Product Name] project at the time of the release to User Acceptance Testing (UAT) .

2. Defect Analysis

This report shows the number of resolved or closed bugs at each severity lever and how they were resolved.

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Subtotal
By Design	11	2	3	2	18
Duplicate	1	3	4	0	8
External	3	5	0	0	8
Fixed	12	2	5	22	41
Not Reproduced	0	1	0	0	1
Skipped	0	0	1	2	3
Won't Fix	0	4	1	1	7
Totals	27	17	14	27	86

3. Test Case Analysis

This report show the number of test cases that have passed, failed and untested.

Section	Total Cases	Not Tested	Fail	Pass
Print	8	0	0	8
Client Application	49	0	0	49
Security	4	0	0	4
Outsource Shipping	4	0	0	4
Exception Reporting	11	0	0	11
Final Report Output	2	0	0	2
Version Control	1	0	0	1

9. RESULTS

9.1 Performance Metrics

[illegible]

10. ADVANTAGES & DISADVANTAGES

Advantages :

- It is a cost-effective way of getting several people from different locations to attend meetings and conferences.
- It enables employees from across the world to communicate with each other 24×7 and share ideas or solve problems quickly.

Disadvantages :

- Also accuracy depends upon distance between camera and object.
- It takes a lot of time to listen, speak, read, or write to someone.

11. CONCLUSION

The proposed communication system between Deaf and Dumb people and ordinary people are aiming for it when bridging the communication gap between two societies. It provides complete two - sided communication in an efficient manner between the disabled and the normal person.

For communication between deaf person and a second person, a mediator is required to translate sign language of deaf person. But a mediator is required to know the sign language used by deaf person. But this is not always possible since there are multiple sign languages for multiple languages.

So to understand all sign languages, Hand gestures of deaf peoples by normal peoples this system is proposed.

12. FUTURE SCOPE

The speech-to-text and text-to-speech technologies helped those people who had difficulties in communicating or expressing their feelings to the normal people.

This reduces the communication gap between the normal people and the specially abled people.

Using image pre-processing and Artificial Intelligence it is easy to understand the context of objects and clearly explains it to the people who use it for communication.

13. APPENDIX

Source Code :

```
1  import cv2
2
3  video = cv2.VideoCapture(0)
4
5  while True:
6      ret, frame = video.read()
7      cv2.imshow("Frame", frame)
8      k = cv2.waitKey(1)
9      if k == ord('q'):
10         break
11
12 video.release()
13 cv2.destroyAllWindows()
```

```
1  import cv2
2  import numpy as np
3  from tensorflow.keras.models import load_model
4  from tensorflow.keras.preprocessing import image
5
6  class Video(object):
7      def __init__(self):
8          self.video = cv2.VideoCapture(0)
9          self.roi_start = (50, 150)
10         self.roi_end = (250, 350)
11         self.model = load_model('asl_model.h5') # Execute Local Trained Model
12         # self.model = load_model('IBM_Communication_Model.h5') # Execute IBM Trained Model
13         self.index=['A','B','C','D','E','F','G','H','I']
14         self.y = None
15
16     def __del__(self):
17         self.video.release()
18
19     def get_frame(self):
20         ret, frame = self.video.read()
21         frame = cv2.resize(frame, (640, 480))
22         copy = frame.copy()
23         copy = copy[150:150+200, 50:50+200]
24         # Prediction Start
25         cv2.imwrite('image.jpg', copy)
26         copy_img = image.load_img('image.jpg', target_size=(64,64))
27         x = image.img_to_array(copy_img)
28         x = np.expand_dims(x, axis=0)
29         pred = np.argmax(self.model.predict(x), axis=1)
30         self.y = pred[0]
31         cv2.putText(frame, 'The Predicted Alphabet is: '+str(self.index[self.y]), (100, 50), cv2.FONT_HERSHEY_SIMPLEX, 1, (0, 0, 0), 3)
32         ret, jpg = cv2.imencode('.jpg', frame)
33         return jpg.tobytes()
```

```

1  <!DOCTYPE html>
2  <html>
3  <head>
4  <meta name="viewport" content="width=device-width, initial-scale=1">
5  <style>
6  body {font-family: Arial, Helvetica, sans-serif;}
7
8  /* Full-width input fields */
9  input[type=text], input[type=password] {
10     width: 100%;
11     padding: 12px 20px;
12     margin: 8px 0;
13     display: inline-block;
14     border: 1px solid #ccc;
15     box-sizing: border-box;
16 }
17
18 /* Set a style for all buttons */
19 button {
20     background-color: #273298;
21     color: white;
22     padding: 14px 20px;
23     margin: 8px 0;
24     border: none;
25     cursor: pointer;
26     width: 100%;
27 }
28
29 button:hover {
30     opacity: 0.8;
31 }

```

```

33 /* Extra styles for the cancel button */
34 .cancelbtn {
35     width: auto;
36     padding: 10px 18px;
37     background-color: #f44336;
38 }
39
40 /* Center the image and position the close button */
41 .imgcontainer {
42     text-align: center;
43     margin: 24px 0 12px 0;
44     position: relative;
45 }
46
47 img.avatar {
48     width: 40%;
49     border-radius: 50%;
50 }
51
52 .container {
53     padding: 16px;
54 }
55
56 span.psw {
57     float: right;
58     padding-top: 16px;
59 }
60
61 /* The Modal (background) */
62 .modal {
63     display: none; /* Hidden by default */
64     position: fixed; /* Stay in place */
65     z-index: 1; /* Sit on top */
66     left: 0;
67     top: 0;
68     width: 100%; /* Full width */
69     height: 100%; /* Full height */

```

```

70     overflow: auto; /* Enable scroll if needed */
71     background-color: □ rgb(0,0,0); /* Fallback color */
72     background-color: □ rgba(0,0,0,0.4); /* Black w/ opacity */
73     padding-top: 60px;
74 }
75
76 /* Modal Content/Box */
77 .modal-content {
78     background-color: ■ #fefefe;
79     margin: 5% auto 15% auto; /* 5% from the top, 15% from the bottom and centered */
80     border: 1px solid ■ #888;
81     width: 80%; /* Could be more or less, depending on screen size */
82 }
83
84 /* The Close Button (x) */
85 .close {
86     position: absolute;
87     right: 25px;
88     top: 0;
89     color: □ #000;
90     font-size: 35px;
91     font-weight: bold;
92 }
93
94 .close:hover,
95 .close:focus {
96     color: ■ red;
97     cursor: pointer;
98 }
99
100 /* Add Zoom Animation */
101 .animate {
102     -webkit-animation: animatezoom 0.6s;
103     animation: animatezoom 0.6s
104 }

```

```

106 @-webkit-keyframes animatezoom {
107     from {-webkit-transform: scale(0)}
108     to {-webkit-transform: scale(1)}
109 }
110
111 @keyframes animatezoom {
112     from {transform: scale(0)}
113     to {transform: scale(1)}
114 }
115
116 /* Change styles for span and cancel button on extra small screens */
117 @media screen and (max-width: 300px) {
118     span.psw {
119         display: block;
120         float: none;
121     }
122     .cancelbtn {
123         width: 100%;
124     }
125 }
126 </style>
127 </head>
128 <body>
129
130 <h2 style="text-align: center; color: □ rgba(1, 2, 2, 0.874);">REAL TIME COMMUNICATION SYSTEM POWERED BY AI FOR SPECIALLY ABLED</h2>
131
132 <button onclick="document.getElementById('id01').style.display='block'">Login</button>
133
134 <div id="id01" class="modal">
135
136     <form class="modal-content animate" action="/action_page.php" method="post">
137         <div class="imgcontainer">
138             <span onclick="document.getElementById('id01').style.display='none'" class="close" title="Close Modal">&times;</span>
139             
140         </div>
141

```



```

142 <div class="container">
143   <label for="uname"><b>Username</b></label>
144   <input type="text" placeholder="Enter Username" name="uname" required>
145
146   <label for="psw"><b>Password</b></label>
147   <input type="password" placeholder="Enter Password" name="psw" required>
148
149   <button type="submit">Login</button>
150   <label>
151     <input type="checkbox" checked="checked" name="remember"> Remember me
152   </label>
153 </div>
154
155 <div class="container" style="background-color: #f1f1f1">
156   <button type="button" onclick="document.getElementById('id01').style.display='none'" class="cancelbtn">Cancel</button>
157   <span class="psw">Forgot <a href="#">password?</a></span>
158 </div>
159 </form>
160 </div>
161 <!doctype html>
162 <html lang="en">
163 <head>
164   <meta charset="UTF-8">
165   <meta name="viewport"
166     content="width=device-width, user-scalable=no, initial-scale=1.0, maximum-scale=1.0, minimum-scale=1.0">
167   <meta http-equiv="X-UA-Compatible" content="ie=edge">
168   <link rel="stylesheet" href="https://stackpath.bootstrapcdn.com/bootstrap/4.1.3/css/bootstrap.min.css">
169   <link rel="stylesheet" href="style.css">
170   <title>Document</title>
171 </head>
172 <body>
173 <div class="display-cover">
174   <video autoplay></video>
175   <canvas class="d-none"></canvas>
176
177   <div class="video-options">
178     <select name="" id="" class="custom-select">

```

```

179       <option value="">Select camera</option>
180     </select>
181   </div>
182
183   <img class="screenshot-image d-none" alt="">
184
185   <div class="controls">
186     <button class="btn btn-danger play" title="Play"><i data-feather="play-circle"></i></button>
187     <button class="btn btn-info pause d-none" title="Pause"><i data-feather="pause"></i></button>
188     <button class="btn btn-outline-success screenshot d-none" title="ScreenShot"><i data-feather="image"></i></button>
189   </div>
190 </div>
191
192 <script src="https://unpkg.com/feather-icons"></script>
193 <script src="script.js"></script>
194 </body>
195 <html><head>
196 </head><body>
197   <video src="" ></video>
198   <br />
199 <button id='flipCamera'>Flip</button>
200 </body>
201 <script>
202   var front = false;
203   var video = document.querySelector('video');
204   document.getElementById('flipCamera').onclick = function() { front = !front; };
205   var constraints = { video: { facingMode: (front? "user" : "environment"), width: 640, height: 480 } };
206   navigator.mediaDevices.getUserMedia(constraints)
207     .then(function(mediaStream) {
208       video.srcObject = mediaStream;
209       video.onloadedmetadata = function(e) {
210         video.play();
211       };
212     })
213     .catch(function(err) { console.log(err.name + ": " + err.message); })
214 </script></html>
215 </html>

```



```

216 <style>
217 .screenshot-image {
218   width: 150px;
219   height: 90px;
220   border-radius: 4px;
221   border: 2px solid whitesmoke;
222   box-shadow: 0 1px 2px 0 rgba(0, 0, 0, 0.1);
223   position: absolute;
224   bottom: 5px;
225   left: 10px;
226   background: white;
227 }
228
229 .display-cover {
230   display: flex;
231   justify-content: center;
232   align-items: center;
233   width: 70%;
234   margin: 5% auto;
235   position: relative;
236 }
237
238 video {
239   width: 100%;
240   background: rgba(0, 0, 0, 0.2);
241 }
242
243 .video-options {
244   position: absolute;
245   left: 20px;
246   top: 30px;
247 }
248
249 .controls {
250   position: absolute;
251   right: 20px;
252   top: 20px;

```

```

253   display: flex;
254 }
255
256 .controls > button {
257   width: 45px;
258   height: 45px;
259   text-align: center;
260   border-radius: 100%;
261   margin: 0 6px;
262   background: transparent;
263 }
264
265 .controls > button:hover svg {
266   color: white !important;
267 }
268
269 @media (min-width: 300px) and (max-width: 400px) {
270   .controls {
271     flex-direction: column;
272   }
273
274   .controls button {
275     margin: 5px 0 !important;
276   }
277 }
278
279 .controls > button > svg {
280   height: 20px;
281   width: 18px;
282   text-align: center;
283   margin: 0 auto;
284   padding: 0;
285 }
286
287 .controls button:nth-child(1) {
288   border: 2px solid #1a12b3;
289 }

```

```

291 .controls button:nth-child(1) svg {
292 |   color: #2b128e;
293 | }
294
295 .controls button:nth-child(2) {
296 |   border: 2px solid #008496;
297 | }
298
299 .controls button:nth-child(2) svg {
300 |   color: #008496;
301 | }
302
303 .controls button:nth-child(3) {
304 |   border: 2px solid #0048b5;
305 | }
306
307 .controls button:nth-child(3) svg {
308 |   color: #0f0a5b;
309 | }
310
311 .controls > button {
312 |   width: 45px;
313 |   height: 45px;
314 |   text-align: center;
315 |   border-radius: 100%;
316 |   margin: 0 6px;
317 |   background: transparent;
318 | }
319
320 .controls > button:hover svg {
321 |   color: rgb(75, 173, 230);
322 | }
323 </style>
324
325 <script>
326 // Get the modal
327 var modal = document.getElementById('id01');

```

```

329 // When the user clicks anywhere outside of the modal, close it
330 window.onclick = function(event) {
331 |   if (event.target == modal) {
332 |     modal.style.display = "none";
333 |   }
334 | }
335 feather.replace();
336
337 const controls = document.querySelector('.controls');
338 const cameraOptions = document.querySelector('.video-options>select');
339 const video = document.querySelector('video');
340 const canvas = document.querySelector('canvas');
341 const screenshotImage = document.querySelector('img');
342 const buttons = [...controls.querySelectorAll('button')];
343 let streamStarted = false;
344
345 const [play, pause, screenshot] = buttons;
346
347 const constraints = {
348 |   video: {
349 |     width: {
350 |       min: 1280,
351 |       ideal: 1920,
352 |       max: 2560,
353 |     },
354 |     height: {
355 |       min: 720,
356 |       ideal: 1080,
357 |       max: 1440
358 |     },
359 |   }
360 | };
361 </script>
362 <script>
363 const getCameraSelection = async () => {
364 |   const devices = await navigator.mediaDevices.enumerateDevices();

```

```

365     const videoDevices = devices.filter(device => device.kind === 'videoinput');
366     const options = videoDevices.map(videoDevice => {
367       return `<option value="${videoDevice.deviceId}">${videoDevice.label}</option>`;
368     });
369     cameraOptions.innerHTML = options.join('');
370   });
371
372 </script>
373 <script>
374
375   play.onclick = () => {
376     if (streamStarted) {
377       video.play();
378       play.classList.add('d-none');
379       pause.classList.remove('d-none');
380       return;
381     }
382     if ('mediaDevices' in navigator && navigator.mediaDevices.getUserMedia) {
383       const updatedConstraints = {
384         ...constraints,
385         deviceId: {
386           exact: cameraOptions.value
387         }
388       };
389       startStream(updatedConstraints);
390     }
391   };
392
393   const startStream = async (constraints) => {
394     const stream = await navigator.mediaDevices.getUserMedia(constraints);
395     handleStream(stream);
396   };
397
398   const handleStream = (stream) => {
399     video.srcObject = stream;
400     play.classList.add('d-none');
401     pause.classList.remove('d-none');

```

```

402     screenshot.classList.remove('d-none');
403     streamStarted = true;
404   };
405
406   getCameraSelection();
407
408   cameraOptions.onchange = () => {
409     const updatedConstraints = {
410       ...constraints,
411       deviceId: {
412         exact: cameraOptions.value
413       }
414     };
415     startStream(updatedConstraints);
416   };
417
418   const pauseStream = () => {
419     video.pause();
420     play.classList.remove('d-none');
421     pause.classList.add('d-none');
422   };
423
424   const doScreenshot = () => {
425     canvas.width = video.videoWidth;
426     canvas.height = video.videoHeight;
427     canvas.getContext('2d').drawImage(video, 0, 0);
428     screenshotImage.src = canvas.toDataURL('image/webp');
429     screenshotImage.classList.remove('d-none');
430   };
431
432   pause.onclick = pauseStream;
433   screenshot.onclick = doScreenshot;
434 </script>
435
436 </body>
437 </html>

```



```

1 <!DOCTYPE html>
2 <html lang="en">
3
4 <head>
5   <meta charset="utf-8">
6   <meta name="viewport" content="width=device-width, initial-scale=1.0, shrink-to-fit=no">
7   <title>SmartBridge WebApp VideoTemplate</title>
8   <link rel="stylesheet" href="https://cdn.jsdelivr.net/npm/bootstrap@5.1.3/dist/css/bootstrap.min.css">
9   <link rel="stylesheet" href="https://use.fontawesome.com/releases/v5.12.0/css/all.css">
10  <link rel="stylesheet" href="assets/css/Banner-Heading-Image.css">
11  <link rel="stylesheet" href="assets/css/Navbar-Centered-Brand.css">
12  <link rel="stylesheet" href="assets/css/styles.css">
13 </head>
14
15 <body style="background: #394348;">
16   <nav class="navbar navbar-light navbar-expand-md py-3" style="background: #212529;">
17     <div class="container">
18       <div></div><a class="navbar-brand d-flex align-items-center" href="#"><span
19         class="bs-icon-sm bs-icon-rounded bs-icon-primary d-flex justify-content-center align-items-center me-2 bs-icon"><i
20           class="fas fa-flask"></i></span><span style="color: #212529;">Real-time Communication
21         System Powered By AI&nbsp;  For Specially Abled</span></a>
22     </div>
23   </nav>
24
25   <section>
26     <div class="d-flex flex-column justify-content-center align-items-center">
27       <div class="d-flex flex-column justify-content-center align-items-center" id="div-video-feed"
28         style="width: 640px; height: 480px; margin: 10px; min-height: 480px; min-width: 640px; border-radius: 10px; border: 4px dashed #212529;">
29         
31       </div>
32     </div>
33     <div class="d-flex flex-column justify-content-center align-items-center" style="margin-bottom: 10px;"><button
34       class="btn btn-info" type="button" data-bs-target="#modal-1" data-bs-toggle="modal">Quick Reference
35     </button> ASI Alphabets</div>
36   </section>
37 </body>

```

```

38 <div class="container">
39   <div class="accordion text-white" role="tablist" id="accordion-1">
40     <div class="accordion-item" style="background: #333741;">
41       <h2 class="accordion-header" role="tab"><button class="accordion-button" data-bs-toggle="collapse"
42         data-bs-target="#accordion-1 .item-1" aria-expanded="true"
43         aria-controls="accordion-1 .item-1"
44         style="background: #394348; color: #212529;">About The Project</button></h2>
45       <div class="accordion-collapse collapse show item-1" role="tabpanel" data-bs-parent="#accordion-1">
46         <div class="accordion-body">
47           <p class="mb-0">Artificial Intelligence has made it possible to handle our daily activities
48             in new and simpler ways. With the ability to automate tasks that normally require human
49             intelligence, such as speech and voice recognition, visual perception, predictive text
50             functionality, decision-making, and a variety of other tasks, AI can assist people with
51             disabilities by significantly improving their ability to get around and participate in
52             daily activities.<br><br>Currently, Sign Recognition is available <strong>only for
53             alphabets A-I</strong> and not for J-Z, since J-Z alphabets also require Gesture
54             Recognition for them to be able to be predicted correctly to a certain degree of
55             accuracy.</p>
56         </div>
57       </div>
58     </div>
59     <div class="accordion-item" style="background: #333741;">
60       <h2 class="accordion-header" role="tab"><button class="accordion-button collapsed"
61         data-bs-toggle="collapse" data-bs-target="#accordion-1 .item-2" aria-expanded="false"
62         aria-controls="accordion-1 .item-2"
63         style="background: #394348; color: #231241;">Developed By</button></h2>
64       <div class="accordion-collapse collapse item-2" role="tabpanel" data-bs-parent="#accordion-1">
65         <div class="accordion-body">
66           <p class="mb-0">Students at VIT-Bhopal University during SmartBridge AI Externship
67             Program.<br><br>1. <strong>Mirlov Deb</strong> 198CG10067<br>2.
68             <strong>Kushagra</strong> 198CG10025<br>3. <strong>Kartik Dhasmana</strong> 198CG10002
69           </p>
70         </div>
71       </div>
72     </div>
73   </div>
74 </div>

```

```

75 </section>
76 <div class="modal fade" role="dialog" tabindex="-1" id="modal-1">
77   <div class="modal-dialog" role="document">
78     <div class="modal-content">
79       <div class="modal-header">
80         <h4 class="modal-title">American Sign Language - Alphabets</h4><button type="button"
81           class="btn-close" data-bs-dismiss="modal" aria-label="Close"></button>
82       </div>
83       <div class="modal-body"></div>
84       <div class="modal-footer"><button class="btn btn-secondary" type="button"
85         data-bs-dismiss="modal">Close</button></div>
86     </div>
87   </div>
88 </div>
89 <script src="https://cdn.jsdelivr.net/npm/bootstrap@5.1.3/dist/js/bootstrap.bundle.min.js"></script>
90 </body>
91
92 </html>

```

```

1  .bs-icon {
2    --bs-icon-size: .75rem;
3    display: flex;
4    flex-shrink: 0;
5    justify-content: center;
6    align-items: center;
7    font-size: var(--bs-icon-size);
8    width: calc(var(--bs-icon-size) * 2);
9    height: calc(var(--bs-icon-size) * 2);
10   color: var(--bs-primary);
11 }
12
13 .bs-icon-xs {
14   --bs-icon-size: 1rem;
15   width: calc(var(--bs-icon-size) * 1.5);
16   height: calc(var(--bs-icon-size) * 1.5);
17 }
18
19 .bs-icon-sm {
20   --bs-icon-size: 1rem;
21 }
22
23 .bs-icon-md {
24   --bs-icon-size: 1.5rem;
25 }
26
27 .bs-icon-lg {
28   --bs-icon-size: 2rem;
29 }
30
31 .bs-icon-xl {
32   --bs-icon-size: 2.5rem;
33 }
34
35 .bs-icon.bs-icon-primary {
36   color: var(--bs-white);
37   background: var(--bs-primary);

```

```

38 }
39
40 .bs-icon.bs-icon-primary-light {
41   color: var(--bs-primary);
42   background: rgba(var(--bs-primary-rgb), .2);
43 }
44
45 .bs-icon.bs-icon-semi-white {
46   color: var(--bs-primary);
47   background: rgba(255, 255, 255, .5);
48 }
49
50 .bs-icon.bs-icon-rounded {
51   border-radius: .5rem;
52 }
53
54 .bs-icon.bs-icon-circle {
55   border-radius: 50%;
56 }

```

REAL TIME COMMUNICATION SYSTEM POWERED BY AI FOR SPECIALLY ABLED

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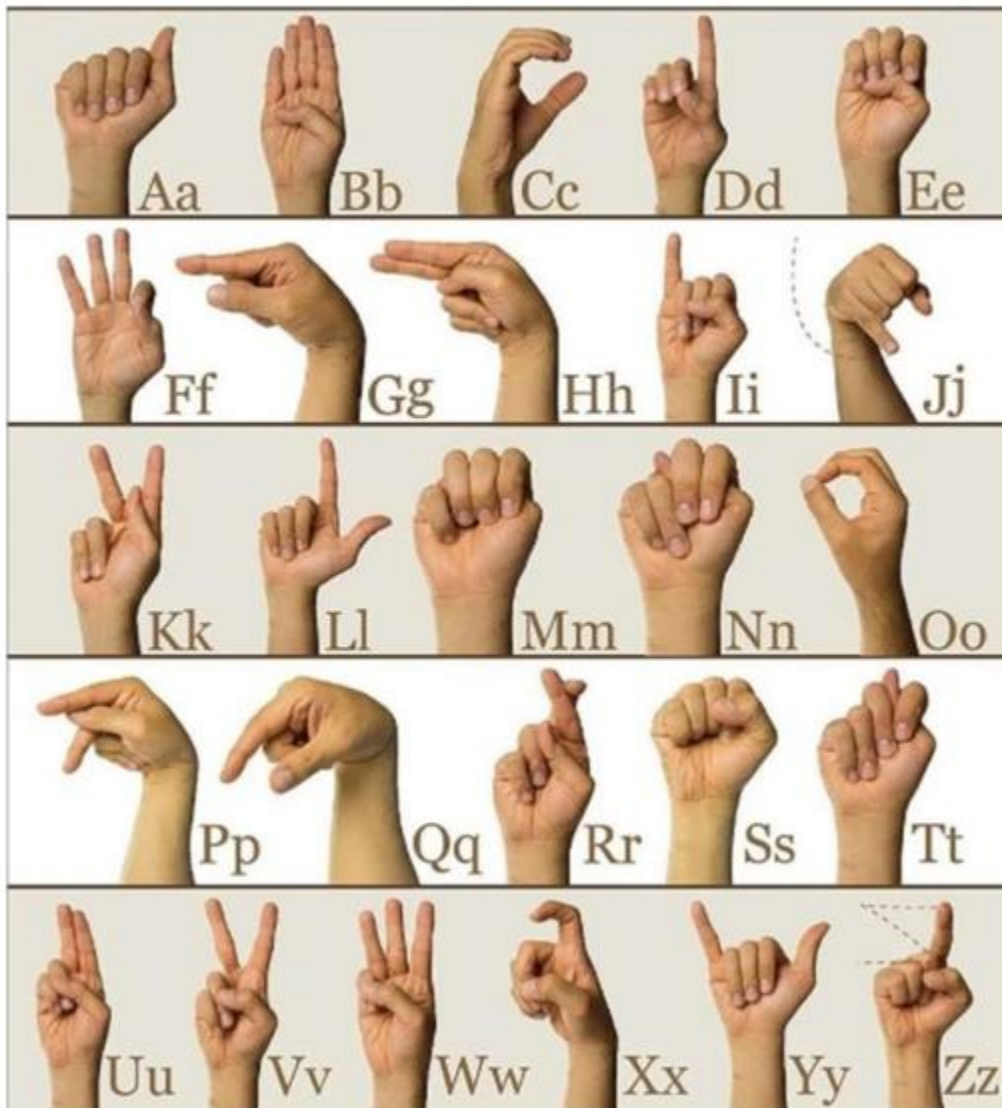
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GITHUB LINK :

<https://github.com/IBM-EPBL/IBM-Project-47296-1660798042>