

ANNAI COLLEGE OF ENGINEERING AND TECHNOLOGY

Kovilacheri in kumbakonam

**REAL-TIME COMMUNICATION SYSTEM POWERED
BY AI FOR SPECIALLY ABLED**

PROJECT REPORT

Submitted by

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ABSTRACT

Deaf and mute people use sign language to communicate. Unlike acoustically conveyed sound patterns, sign language uses hand gestures, facial expressions, body language and manual communication to convey thoughts. Due to the considerable time required in learning Sign Language, people find it difficult to communicate with specially-abled people, creating a communication gap. Hence conventionally, people face problems in recognizing sign language. Moreover, different countries have their respective form of sign gesture communication which results in non-uniformity. The ISL (Indian Sign Language) used in India is largely different from the American Sign Language used in the US, mostly because of the difference in culture, geographical and historical context. Somewhere between 138 and 300 different types of sign language are currently being used throughout the world. Sign language structure varies spatially and temporally. We have identified these as a major barrier in communication with a significant part of society. And hence, we propose to design a system that recognizes different signs and conveys the information to people. The component of any sign language consists of hand shape, motion, and place of articulation. When combined, these three components (together with palm orientation) uniquely determine the meaning of the manual sign. For sign language identification, sensor-based and vision-based methods are used. In vision-based gesture recognition technology, a camera reads the movements of the human body, typically hand movements and uses these gestures to interpret sign language, whereas in sensor-based methods, real-time

hand and finger movements can be monitored using the leap motion sensor. We aim at developing a scalable project where we will be considering different hand gestures to recognize the letters and words. We plan to use different deep learning models to predict the sign. This may be developed as a desktop or mobile application to enable specially abled people to communicate easily and effectively with others. However, this project can later be extended to capture the whole vocabulary of ASL (American Sign Language) through manual and non-manual signs.

Keywords: Sign language, ASL, ISL, Dynamic hand gesture recognition

1.INTRODUCTION

1.1Project Overview

Real-time communications (RTC) is any mode of telecommunications in which all users can exchange information instantly or with negligible latency or transmission delays. In RTC, there is always a direct path between the source and the destination. Although the link might contain several intermediate nodes, the data goes from source to destination without being stored in between them. In contrast, asynchronous or time shifting communications, such as email and voicemail, always involve some form of data storage between the source and the destination. In these cases, there is an anticipated delay between the transmission and receipt of the information.

1.2 PROBLEM STATEMENT

The Deaf and mute community can only communicate using sign language. Sign language involves simultaneously combining hand shapes, orientations, gestures and movement of the hands, arms, or body to express the speaker's thoughts. Because of cultural, geographic and historical differences, there exists over 300 different types of sign languages around the world. The ISL (Indian Sign Language) used in India is very different from the American Sign Language used in the United States. This causes inconsistency of sign languages around the world. Moreover, learning sign language requires significant amount of time and effort. This makes it difficult for the conventional world to learn and hence interact with the deaf and mute community. According to a recent study, out of every thousand kids born, 2 to 3 of them are deaf or hard-of-hearing, and, as degrees of hearing loss go, there are 16 to 30 times more children who are identified as Deaf (having a Profound 91+dB hearing loss) than hard-of-hearing. For those deaf or hard of hearing children, only 10% of parents & family learn sign language to communicate with them. We identify this as a major barrier in communicating with a significant part of the society. 1.2 Purpose Real-time communication (RTC) refers to any communication that happens between two (or more) individuals in real-time – with minimal latency and without transmission delays. Some examples of real-time communication include landline phones, mobile calls, instant messaging, VoIP, and video conferencing.

OBJECTIVE AND MOTIVATION

The objective of our project is to bridge the gap and ensure the inclusion of deaf and mute community into the conventional society meanwhile ensuring an easy and effective mode of communication. We aim at designing a real time system that recognizes the sign language and expresses the same in an easy language, like English. Currently, extensive work has been done on American sign language recognition, but Indian sign language differs significantly from American sign language. ISL uses two hands for communicating (20 out of 26) whereas ASL uses single hand for communicating. Using both hands often lead to obscurity of features due to overlapping of hands. In addition to this, lack of datasets and variance in sign language with locality has resulted in restrained efforts in ISL gesture detection. Our 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 mute people communicate faster and easier

with outer world, but also provide a boost in developing autonomous systems for understanding and aiding them.

Communication between Deaf and Mute People and Normal People Chat applications have become a powerful media that assist people to communicate in different languages with each other. There are lots of chat applications that are used different people in different languages but there is not such a chat application that has facilitated to communicate with sign languages. The developed system is based on Sinhala Sign language. The system has included four main components as text messages are converted to sign messages, voice messages are converted to sign messages, sign messages are converted to text messages and sign messages are converted to voice messages. Google voice recognition API has used to develop speech character recognition for voice messages. The system has been trained for the speech and text patterns by using some text parameters and signs of Sinhala Sign language is displayed by emojis. Those emojis and signs that are included in this system will bring the normal people closer to the disabled people. This is a 2-way communication system, but it uses pattern of gesture recognition which is not very reliable in getting appropriate output.

Intelligent Sign Language Recognition

Using Image Processing Computer recognition of sign language is an important research problem for enabling communication with hearing impaired people. This project introduces an efficient and fast algorithm for identification of the number of fingers opened in a gesture representing an alphabet of the Binary Sign Language. The system does not require the hand to be perfectly aligned to the camera. The project uses image processing system to identify, especially English alphabetic sign language used by the deaf people to communicate. The basic objective of this project is to develop a computer based intelligent system that will enable dumb people significantly to communicate with all other people using their natural hand gestures. The idea consisted of designing and building up an intelligent system using image processing, machine learning and artificial intelligence concepts to take visual inputs of sign language's hand gestures a generate easily recognizable form of outputs. Hence the objective of this project is to develop an intelligent system which can act as a translator between the sign language and the spoken language dynamically and can make the communication between people with hearing impairment and normal people both effective and efficient. The system is we are implementing for Binary sign language, but it can detect any sign language with prior image processing.

Sign Language Recognition Using Image Processing One of the major drawbacks of our society is the barrier that is created between disabled or handicapped persons and the normal person. Communication is the only medium by which we can share our thoughts or convey the message but for a person with disability (deaf and mute) faces difficulty in communication with normal person. For many deaf and dumb people, sign language is the basic means of communication. Sign language recognition (SLR) aims to interpret sign languages automatically by a computer in order to help the deaf communicate with hearing society conveniently. Our aim is to design a system to help the person who trained the hearing impaired to communicate with the rest of the world using sign language or hand gesture recognition techniques. In this system, feature detection and feature extraction of hand 23 gesture is done with the help of SURF algorithm using image processing. All this work is done using MATLAB software. With the help of this algorithm, a person can easily train a deaf and mute

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. Communication 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.

2.2 REFERENCES

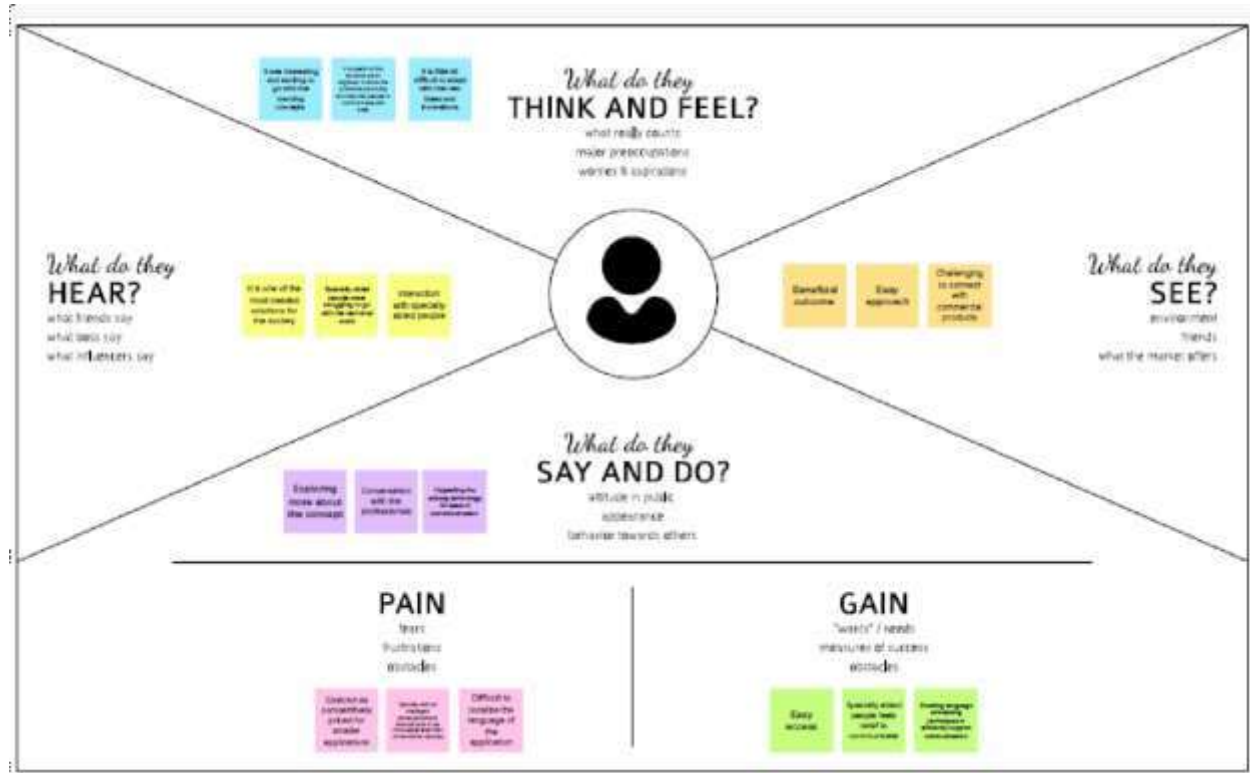
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2.3 Problem Statement Definition

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 deafmute 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. 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.IDEATION AND PROPOSED SOLUTION

3.1 Empathy Map Canvas



3.2 Ideation & brainstorming



3.3 PROPOSED SOLUTION

The motive of our application is to make deaf-dumb people communicate easily with the people by the help of real-time system.

S.No	Parameter	Description	
1	Problem statement (problem to be solved)	□	To solve the issues of deaf-dumb people to communicate with the people to make them feel confident
2	Idea/Solution description	□	Converting sign language into voice and text in the desired language (two-way communication) using Convolutional Neural Network technology.
3	Novelty/Uniqueness	□	Upgrading our solution by implementing an alert system using Big Panda algorithm for improvement
4	Social Impact/Customer Satisfaction	• •	To reduce the risk of losing their lives. It increases the scope for career development. It will smash all the barriers and will help to enhance their skills in a positive manner.
5	Business Model(financial Benefit)	• •	We will collaborate with multi deaf-dumb organizations to out spread the application. Here we give most of the basic features at free cost but they have to pay if they need more advanced features.
6	Scalability of Solution	• •	It has very less complexity for the user Encoding the errors and decoding with better accuracy.

3.3PROBLEM SOLUTION FIT

Define CS, fit into CC	1. CUSTOMER SEGMENT(S) <small>Who is your customer? i.e. working parents of 0-5 y.o. kids</small> Specially abled persons.	6. CUSTOMER CONSTRAINTS <small>What constraints prevent your customers from taking action or limit their choice of solutions? (i.e. spending power, budget, no cash, network connection, available devices)</small> Implanted electronic medical device that can produce useful hearing sensation by electrically stimulating nerves inside the inner ear.	5. AVAILABLE SOLUTIONS <small>Which solutions are available to the customers when they face the problem? What do they need to get the job done? What have they tried in the past? What pros & cons do these solutions have? (i.e. pen and paper is an alternative to digital notetaking)</small> The first ever approach to sign language it has only 6 sign gestures detection. As AI takes an important role in communication and interaction, the use of this technology enables individuals with disabilities to access information much easier, all just by speaking to their devices.	Explore AS, differentiate
	2. JOBS-TO-BE-DONE / PROBLEMS <small>Which jobs/tasks/done (or problems) do you address for your customer? There could be more than one, explore different sides</small> Deaf and dumb people couldn't able to convey their messages to the normal people easily. Deaf people cannot hear the words as others speak and dumb people cannot express their feelings by words. Concentrate on making their communication much easier and live a normal life.	9. PROBLEM ROOT CAUSE <small>What is the real reason that this problem exists? What is the back-story behind the need to do this job? (i.e. customers have to do it because of the change in regulations)</small> Disabilities affect the entire family. Meeting the complex needs of a person with a disability can put families under a great deal of stress — emotional, financial, and sometimes even physical. However, finding resources, knowing what to expect, and planning for the future can greatly improve overall quality of life.	7. BEHAVIOUR <small>What does your customer do to address the problem and get the job done? Is directly related? Find the right solar panel installer, calculate usage and benefits, indirectly associated, customers spend free time of volunteering work (i.e. Greenpeace)</small> In our device, there's an option called problem detection display in which our customer can able to see the type of problem occurs & solution will be displayed.	
Identify strong TR & EM	3. TRIGGERS <small>What triggers customers to act? (i.e. seeing their neighbour installing solar panels, reading about a more efficient solution in the news)</small> By comparing normal people, Specially Abled people should depend on others and want to live their life independently like other people.	10. YOUR SOLUTION <small>If you are working on an existing business, write down your current solution first, fill in the canvas, and check how much it fits reality. If you are working on a new business proposition, then keep it blank until you fill in the canvas and come up with a solution that fits within customer limitations, solves a problem and matches customer behaviour.</small> Facial recognition, voice recognition and predictive texting tools allows people who have difficulties in speaking to communicate more easily using AI. 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 <small>What kind of actions do customers take online? Extract online channels from #7</small> Advertise on online with influencers to test the product and promote it also on social medias.	Identify strong TR & EM
	4. EMOTIONS: BEFORE / AFTER <small>How do customers feel when they face a problem or a job and afterwards? i.e. love, pleasure + confident, in control - use it in your communication strategy & design</small> BEFORE: It is very difficult to convey the message to normal people. AFTER: They overcome their reluctance to have communication with normal people.		8.2 OFFLINE <small>What kind of actions do customers take offline? Extract offline channels from #7 and use them for customer development.</small>	

4 REQUIREMENT ANALYSIS

4.1 FUNCTIONAL REQUIREMENT

- System is presented as black box
 - Hearing impaired is the person that performs the signs
 - Normal hearing is the passive user of the system
- The System Requirements Can Be Specified
1. Hearing impaired person should be able to perform sign that represent digit number
 2. Hearing impaired person should be able to perform sign that represent alphabet letter 29
 3. Hearing impaired person should be able to perform sign that represent word
 4. Hearing impaired person should be able to perform sign that represent sentence
 5. Hearing impaired person should be able to see the translation of sign to text
 6. Hearing impaired person should be able to change the component (number/alphabet or word/sentence) for which translation to speech is provided

NORMAL FLOW

1. User comes in front of camera and performs the alphabet letter
2. System analyzes the performed sign
3. System shows the sign meaning as text and speech

ALTERNATIVE FLOWS

System indicates that user is not within field of view of Kinect

1. System shows that user is not detected
2. User enters the field of view
3. System shows that user is detected Sign not recognized

1. System does not react to indicate that sign was not recognized
2. User performs again the alphabet letter until it is recognized

Enabling speech for this component:

1. Enable speech component

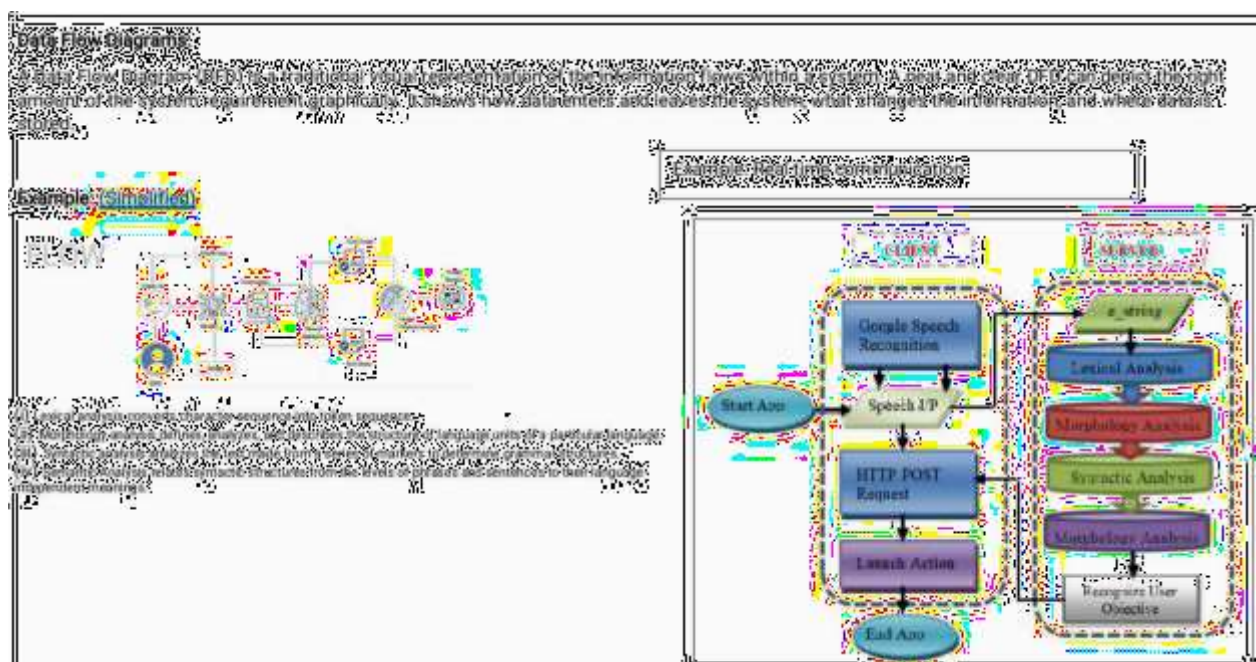
4.2 NON FUNCTIONAL REQUIREMENT

FRNo.	Non-FunctionalRequirement	Description
NFR-1	Usability	The designed system is easy to use for speciallyabledpersonsasitisportableand platformindependent.
NFR-2	Security	Convertedinformationusingsignsintospeechis accessed only by the user.
NFR-3	Reliability	Systemistestedwithlargenumberofdataand Providesinsight into issues.
NFR-4	Performance	QuickLaunchtimeofapplicationandfasterinconverting signs into speech
NFR-5	Availability	Providesautomaticrecoveryand Useraccess.
NFR-6	Scalability	Standard network condition the device shouldconvertinformationwithinsecond.

5 PROJECT DESIGN

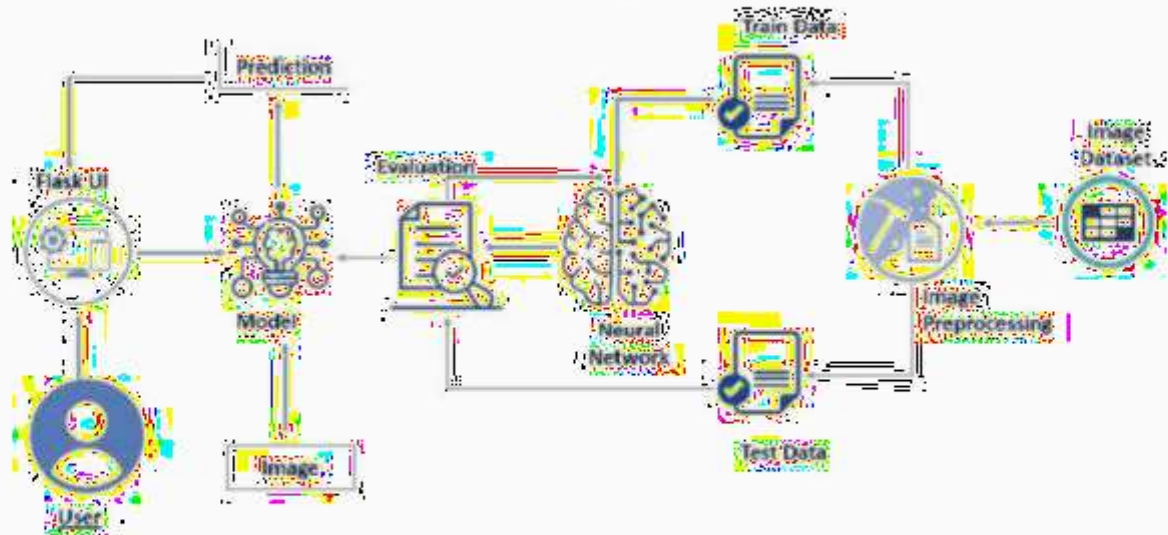
5.1 DATA FLOW DIAGRAM

A data flow diagram is a traditional visual representation of the information flow within a system. It shows how data enters and leaves the system. It uses defined symbols like rectangles, circles and arrows, plus short text labels, to show data inputs, outputs, storage points and the routes between each destination.



User Stories						
User Type	Functional Requirement (User)	User Story Number	User Story / Task	Acceptance Criteria	Priority	Release
Developer	Registration	US001	As a user, I want to register my account by entering my email address and password.	Can register my account successfully.	High	Version 1.0
		US002	As a user, I want to login to my account using my email address and password.	Can login to my account successfully.	High	Version 1.0
		US003	As a user, I want to reset my password if I forget it.	Can reset my password successfully.	Medium	Version 1.0
		US004	As a user, I want to create my account using my phone number and a verification code.	Can create my account successfully using phone number.	Medium	Version 1.0
	Dashboard	US005	As a user, I can create my account in the dashboard.	Can create my account successfully in the dashboard.	High	Version 1.0
Admin	Dashboard	US006	As an admin, I can manage the application content.	Can manage application content successfully.	High	Version 1.0
		US007	As an admin, I can manage the application content.	Can manage application content successfully.	High	Version 1.0
		US008	As an admin, I can manage the application content.	Can manage application content successfully.	High	Version 1.0
		US009	As an admin, I can manage the application content.	Can manage application content successfully.	High	Version 1.0
	Report	US010	As an admin, I can generate reports of the application.	Can generate reports of the application successfully.	Medium	Version 1.0
	Dashboard	US011	As a user, I can create my account in the dashboard.	Can create my account successfully in the dashboard.	High	Version 1.0
Admin	Dashboard	US012	As an admin, I can manage the application content.	Can manage application content successfully.	High	Version 1.0
Admin	Report	US013	As an admin, I can generate reports of the application.	Can generate reports of the application successfully.	High	Version 1.0
Admin	Report	US014	As an admin, I can generate reports of the application.	Can generate reports of the application successfully.	High	Version 1.0

5.2 SOLUTION AND TECHNICAL ARCHITECTURE

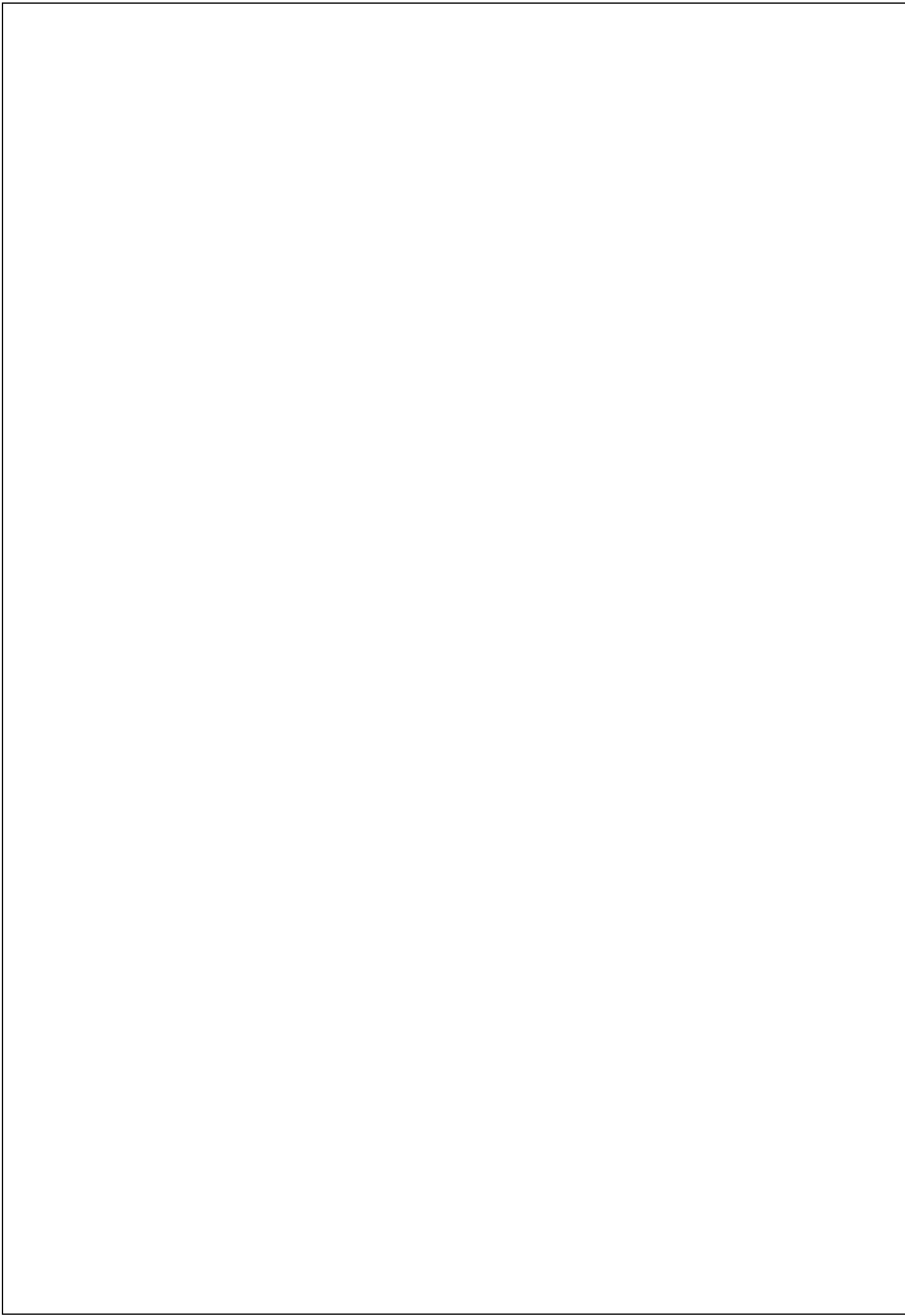


SOLUTION ARCHITECTURE



6 SPRINT DELIVERY PLAN

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.	2	High	Sivaranjani T Privitha K Sukitha M Bhuvamneswari M
Sprint-1	Registration	USN-2	As a user, I will receive confirmation email once I have registered for the application	1	High	Sivaranjani T Sugitha M Bhuvaneshwari M Privitha k
Sprint-2	Registration	USN-3	As a user, I can register for the application through phone number	2	Medium	Sivaranjani T Sugitha M Bhuvamneswari M Privitha K
Sprint-2	User interface	USN-4	Professional responsible for user requirements & needs	2	Medium	Privitha K Sugitha M Bhuvaneshwari m Sivaranjani T
Sprint-3	Login	USN-5	As a user, I can log into the application by entering email & password	1	High	Sivaranjani T Privitha k Sugitha M Bhuvaneshwari M
Sprint-3	Dashboard	USN-6	As a user, I must receive any updates or pop ups in my dashboard	2	High	Bhuvaneshwari M Sivaranjani T Sugitha M Privitha K
Sprint-4	Details	USN-7	As a user, I should get notification about the progress and any updates via emails	1	Medium	Privitha k Bhuvamneswari M Sugitha M Sivaranjani T
Sprint-4	Privacy	USN-8	The developed application should be secure for the users	2	High	Sugitha M Bhuvaneshwari m Sivaranjani t Privitha k



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

Model Building

```
from keras.preprocessing.image import ImageDataGenerator

model.add(Dense(128, activation='relu'))
model.add(Dense(64, activation='relu'))

print(model.summary())
model.compile(optimizer='adam',
              loss='binary_crossentropy',
              metrics=['accuracy'])

# Training the model
train_data_gen = ImageDataGenerator(rescale=1/255, zoom_range=(0.9, 1.1),
                                     rotation_range=10, shear_range=0.1,
                                     width_shift_range=0.1, height_shift_range=0.1,
                                     fill_mode='nearest')
test_data_gen = ImageDataGenerator(rescale=1/255)

# Training the model
history = model.fit(train_data_gen.flow(train_data, train_labels,
                                       batch_size=32),
                    test_data_gen.flow(test_data, test_labels,
                                       batch_size=32),
                    validation_data=(test_data, test_labels),
                    epochs=10,
                    verbose=1)

print('Test accuracy: %f' % history.history['val_accuracy'])
print('Test loss: %f' % history.history['val_loss'])

# Save the model
model.save('model.h5')
```

```
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout, Flatten, Input

# Create the model
model = Sequential()

# Add the input layer
model.add(Dense(128, activation='relu'))

# Add the hidden layers
model.add(Dense(64, activation='relu'))
model.add(Dense(32, activation='relu'))

# Add the output layer
model.add(Dense(1, activation='sigmoid'))

# Compile the model
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
```


Model Building

Adding The Pooling Layer

```
10: type tensorflow.keras.preprocessing.image.ImageDataGenerator

11: # create model to be
type:keras.models.Sequential
type:keras.layers.MaxPooling2D

12: # define input image
image = np.array([
    [1, 4, 5, 2],
    [4, 5, 2, 5],
    [1, 4, 5, 2]])

image = image.reshape(1, 4, 4)

13: # define model (combining what is single maxpooling layer
model = Sequential()
    MaxPooling2D(pool_size = 2, strides = 2))

# generate partial output
output = model.predict(image)

14: # print output image
output = np.squeeze(output)
print(output)

15: # training images
train_images = ImageDataGenerator(rescale=1/255, zoom_range=(1, horizontal_flip=True, vertical_flip=True))
test_images = ImageDataGenerator(rescale=1/255)

16: # training dataset
x_train=train_images.flow_from_directory('dataset/train', target_size=(64,64), class_mode='categorical', batch_size=64)
y_train=train_images.flow_from_directory('dataset/train', target_size=(64,64), class_mode='categorical', batch_size=64)
x_test=test_images.flow_from_directory('dataset/test', target_size=(64,64), class_mode='categorical', batch_size=64)

17: # training dataset
train_images = ImageDataGenerator(rescale=1/255, zoom_range=(1, horizontal_flip=True, vertical_flip=True))
test_images = ImageDataGenerator(rescale=1/255)

18: # training details
x_train=train_images.flow_from_directory('dataset/train', target_size=(64,64), class_mode='categorical', batch_size=64)
y_train=train_images.flow_from_directory('dataset/train', target_size=(64,64), class_mode='categorical', batch_size=64)
x_test=test_images.flow_from_directory('dataset/test', target_size=(64,64), class_mode='categorical', batch_size=64)

19: # print training details
print('train_size: ', len(x_train))
print('test_size: ', len(x_test))

20: # the keras function to training dataset
x_train=train_images

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x_train=train_images

62: # the keras function to training dataset
x_train=train_images

63: # the keras function to training dataset
x_train=train_images

64: # the keras function to training dataset
x_train=train_images

65: # the keras function to training dataset
x_train=train_images

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x_train=train_images

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x_train=train_images

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x_train=train_images

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x_train=train_images

77: # the keras function to training dataset
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x_train=train_images

80: # the keras function to training dataset
x_train=train_images

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x_train=train_images

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x_train=train_images

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x_train=train_images

84: # the keras function to training dataset
x_train=train_images

85: # the keras function to training dataset
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90: # the keras function to training dataset
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92: # the keras function to training dataset
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93: # the keras function to training dataset
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94: # the keras function to training dataset
x_train=train_images

95: # the keras function to training dataset
x_train=train_images

96: # the keras function to training dataset
x_train=train_images

97: # the keras function to training dataset
x_train=train_images

98: # the keras function to training dataset
x_train=train_images

99: # the keras function to training dataset
x_train=train_images
```

Model Building

Model Building

```
from tensorflow.keras.preprocessing import ImageDataGenerator
```

```
train_datagen = ImageDataGenerator(rescale=1/255, rotation_range=10, width_shift_range=0.1, height_shift_range=0.1)
```

```
# Generating data, sometimes we need to shuffle the variables
# e.g. shuffle
# train_datagen = ImageDataGenerator(rescale=1/255, rotation_range=10, width_shift_range=0.1, height_shift_range=0.1, shuffle=True)

# Generating data, sometimes we need to shuffle the variables
# e.g. shuffle
# train_datagen = ImageDataGenerator(rescale=1/255, rotation_range=10, width_shift_range=0.1, height_shift_range=0.1, shuffle=True)
```

```
# Training images
train_datagen = ImageDataGenerator(rescale=1/255, rotation_range=10, width_shift_range=0.1, height_shift_range=0.1)
train_generator = train_datagen.flow_from_directory('data/train', target_size=(256, 256), class_mode='categorical', batch_size=32)
```

```
# Training images
train_datagen = ImageDataGenerator(rescale=1/255, rotation_range=10, width_shift_range=0.1, height_shift_range=0.1)
train_generator = train_datagen.flow_from_directory('data/train', target_size=(256, 256), class_mode='categorical', batch_size=32)
```

```
# Training images
```

```
def compile_model(model, compile_model, model_dir):
    # Compile the model
    # e.g. compile_model(model, compile_model, model_dir)
    # e.g. compile_model(model, compile_model, model_dir)
    # e.g. compile_model(model, compile_model, model_dir)
    # e.g. compile_model(model, compile_model, model_dir)
```

```
# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
```

```
# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
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# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
```

```
# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
# e.g. compile_model(model, compile_model, model_dir)
```

```
# e.g. compile_model(model, compile_model, model_dir)
```

```
# e.g. compile_model(model, compile_model, model_dir)
```

```

174: # Adding layers
175: # Sequential model
176: model = Sequential()
177:
178: # Adding layers
179: model.add(Conv2D(32, (3, 3), activation='relu', input_shape=(64, 64, 3)))
180:
181: model.add(MaxPooling2D(pool_size=(2, 2)))
182: model.add(Flatten())
183:
184: # Adding Dense layers
185: model.add(Dense(128, activation='relu'))
186: model.add(Dense(64, activation='relu'))
187: model.add(Dense(1, activation='sigmoid'))
188:
189: # Compiling the model
190: model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
191:
192: # Loading data from a file
193: # Load data
194: temp = f.read()
195: f.close()
196:
197: code = compile(temp, '', 'exec')
198: exec(code)
199:
200: # Saving the model
201: # Save model to file
202:
203: model.save('model.h5')

```

Model Building

Synthesis

From `keras.preprocessing.image_utils`: `image_utils.load_img`

```
# Training Images
train_images = ImageDataGenerator(rescale=1/255, range=[0, 1], horizontal_flip=True, vertical_flip=False)
# Validation Images
test_images = ImageDataGenerator(rescale=1/255)
```

```
# loading source
x_test = load_data(source_dir, target_dir, loader, shuffle=True)
# saving source
x_test.save(source_dir, target_dir, loader, shuffle=True)
```

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```
# Load Model Using Pickle
import pickle

from sklearn import model_selection
from sklearn.linear_model import LogisticRegression
import pickle
```

[illegible]

```

- model = LinearRegression()
- model.fit(X_train, y_train)
- # Print the model's score
- r_squared = model.score(X_train, y_train)
- print(f"R-squared score: {r_squared}")

- # Print the model's coefficients
- coefficients = model.coef_
- intercept = model.intercept_
- print(f"Coefficients: {coefficients}")
- print(f"Intercept: {intercept}")

```

```
print("Low A-THAN") # Low A-THAN
print("Low A-THAN") # Low A-THAN
```

$$\frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

```

00000000 | 0: the train number is beginning with 0
00000000 | 1: train class begins

```

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Journal of Internal Medicine 255: 105–112

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

```

    @Override
    public void onCreate(Bundle savedInstanceState) {
        super.onCreate(savedInstanceState);
        setContentView(R.layout.activity_main);
        // ...
    }
}

```

```
# adding logistic
model.add(layers.Dense(12, (3, 1), activation='relu', input_shape=(4, 16, 1)))
```

[illegible]

code: additivity

[Adding Google Maps](#)
[Adding Google Web Analytics](#)

```
# Adding Dense Layers
model.add(Dense(100, activation='relu'))
model.add(Dense(100, activation='relu'))
model.add(Dense(10, activation='softmax')
```

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

```
# Fitting the Model
model.fit_generator(train_data_generator, train_data_generator.n_samples_per_epoch, validation_data=test_data_generator, validation_steps=100, epochs=100)
```

```
def load_model():
    """Load the trained model from the saved file path"""
    model = load_model('model.h5')
```

```
def predict_image(image_path):
    """Predict the class for a given image path"""
    # Load the image
    image = load_image(image_path)
    # Preprocess the image
    image = preprocess_image(image)
    # Predict the class
    prediction = model.predict(image)
    # Get the predicted class
    predicted_class = np.argmax(prediction)
    # Print the predicted class
    print(f'Predicted class: {predicted_class}')
    return predicted_class
```

def main():

```
    # Load the model
    model = load_model()
```

Model Building

Importing The Required Model Building Libraries

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
def load_model():
    """Load the trained model from the saved file path"""
    model = load_model('model.h5')
```

```
# Training Dataset
train_data_generator = ImageDataGenerator(rotation_range=30, zoom_range=0.1, horizontal_flip=True, validation_split=0.1)
# Testing Dataset
test_data_generator = ImageDataGenerator(rotation_range=30, zoom_range=0.1, horizontal_flip=True, validation_split=0.1)
```

```
# Training Dataset
train_data_generator.flow_from_directory('data/train', target_size=(224, 224), class_mode='categorical', batch_size=32)
# Testing Dataset
test_data_generator.flow_from_directory('data/test', target_size=(224, 224), class_mode='categorical', batch_size=32)
```

```
def load_model():
    """Load the trained model from the saved file path"""
    model = load_model('model.h5')
```

def main():

```
    # Load the model
    model = load_model()
```

```
def predict_image(image_path):
    """Predict the class for a given image path"""
    # Load the image
    image = load_image(image_path)
    # Preprocess the image
    image = preprocess_image(image)
    # Predict the class
    prediction = model.predict(image)
    # Get the predicted class
    predicted_class = np.argmax(prediction)
    # Print the predicted class
    print(f'Predicted class: {predicted_class}')
    return predicted_class
```

```
# Generating Dataset
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Conv2D, MaxPooling2D, Flatten, Softmax
```

```
def main():
    # Load the model
    model = load_model()
```


Initializing The Model:

```
140:17 from tensorflow.keras.preprocessing.image import ImageDataGenerator

140:18
140:19 spatial_dropout=0.5
140:20 recurrent_dropout=0

140:21
140:22 # Loading Datasets
140:23 train_datagen = ImageDataGenerator(rescale=1/255, zoom_range=(0.9,1.1), horizontal_flip=True, rotation=15, shear=0.1)
140:24 # Loading Datasets
140:25 test_datagen = ImageDataGenerator(rescale=1/255)

140:26
140:27 # Training Dataset
140:28 x_train_datagen = train_datagen.flow_from_directory('content/processedImages/train', target_size=(64,64), class_mode='categorical', batch_size=64)
140:29 # Loading Dataset
140:30 x_test_datagen = test_datagen.flow_from_directory('content/processedImages/test', target_size=(64,64), class_mode='categorical', batch_size=64)

140:31
140:32 # Print the shape of the data
140:33 print('x_train shape:', x_train_datagen.np.ndarray[0].shape)
140:34 print('x_test shape:', x_test_datagen.np.ndarray[0].shape)

140:35
140:36 # Print the shape of the data
140:37 print('x_train shape:', x_train_datagen.np.ndarray[0].shape)
140:38 print('x_test shape:', x_test_datagen.np.ndarray[0].shape)

140:39
140:40 # The class indices in Training Dataset
140:41 y_train_class_indices = x_train_datagen.class_indices

140:42
140:43 # The class indices in Test Dataset
140:44 y_test_class_indices = x_test_datagen.class_indices

140:45
140:46 # Loading Libraries
140:47 from tensorflow.keras.models import Sequential
140:48 from tensorflow.keras.layers import Convolution2D, MaxPooling2D, Flatten, Dense

140:49
140:50 # Creating Model
140:51 model = Sequential()
```

Initializing The Model:

```
140:17 from tensorflow.keras.preprocessing.image import ImageDataGenerator

140:18
140:19 spatial_dropout=0.5
140:20 recurrent_dropout=0

140:21
140:22 # Loading Datasets
140:23 train_datagen = ImageDataGenerator(rescale=1/255, zoom_range=(0.9,1.1), horizontal_flip=True, rotation=15, shear=0.1)
140:24 # Loading Datasets
140:25 test_datagen = ImageDataGenerator(rescale=1/255)

140:26
140:27 # Training Dataset
140:28 x_train_datagen = train_datagen.flow_from_directory('content/processedImages/train', target_size=(64,64), class_mode='categorical', batch_size=64)
140:29 # Loading Dataset
140:30 x_test_datagen = test_datagen.flow_from_directory('content/processedImages/test', target_size=(64,64), class_mode='categorical', batch_size=64)

140:31
140:32 # Print the shape of the data
140:33 print('x_train shape:', x_train_datagen.np.ndarray[0].shape)
140:34 print('x_test shape:', x_test_datagen.np.ndarray[0].shape)

140:35
140:36 # Print the shape of the data
140:37 print('x_train shape:', x_train_datagen.np.ndarray[0].shape)
140:38 print('x_test shape:', x_test_datagen.np.ndarray[0].shape)

140:39
140:40 # The class indices in Training Dataset
140:41 y_train_class_indices = x_train_datagen.class_indices

140:42
140:43 # The class indices in Test Dataset
140:44 y_test_class_indices = x_test_datagen.class_indices

140:45
140:46 # Loading Libraries
140:47 from tensorflow.keras.models import Sequential
140:48 from tensorflow.keras.layers import Convolution2D, MaxPooling2D, Flatten, Dense

140:49
140:50 # Creating Model
140:51 model = Sequential()
```

8. TESTING

8.1 Test cases

Real Time Communication System Powered By AI For Specially Abled

Using the Dataset for Image Data Generation

```
# Import tensorflow.keras.preprocessing.image_utils.ImageDataGenerator

# Training dataset
train_datagen = ImageDataGenerator(rescale=1/255, shear_range=0.2, horizontal_flip=True, vertical_flip=False)
# Testing dataset
test_datagen = ImageDataGenerator(rescale=1/255)

# Training dataset
x_train=train_datagen.flow_from_directory('dataset/train', target_size=(96, 96), class_mode='categorical', batch_size=64)
# Testing dataset
x_test=test_datagen.flow_from_directory('dataset/test', target_size=(96, 96), class_mode='categorical', batch_size=64)

# Create model architecture
model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Conv2D(64, kernel_size=(3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Flatten())
model.add(Dense(1000, activation='relu'))
model.add(Dense(1000, activation='relu'))
model.add(Dense(10))
model.compile(optimizer='adam', loss='categorical_crossentropy')

# Fit the model
history = model.fit_generator(x_train, x_test, epochs=10, validation_data=(x_test, y_test))
```

```

model.add(Dense(1024), activation='relu')
model.add(Dense(1024))

# Adding Dense Layers
model.add(Dense(512), activation='relu')
model.add(Dense(512), activation='relu')
model.add(Dense(512), activation='relu')

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

# Fitting the Model
model.fit_generator(train_steps_per_epoch=10, epochs=10, validation_data=(test, validation_steps=10))

```

Warning: The `libffi` package is deprecated and will be removed in a future version. Please use `ffi` instead.

[illegible]

```

11 # Save the Model
12 model.save('cat_model.h5')
13
14 # Testing the model
15
16 # Load a sample image
17 img = test_image.imread('data/pets/000001.jpg')
18 img = test_image.img_to_array(img).astype('float32')
19
20 # Predict the class
21 predicted_class = model.predict(img)
22 print('Predicted class: %s' % predicted_class)
23
24 # Print the target class
25 target_class = test_image.get_target_class('000001.jpg')
26 print('Target class: %s' % target_class)

```


Real-Time Communication System Powered By AI For Specially Abled

Loading the Dataset & Image Data Generation

```
20 [21]: from tensorflow.keras.preprocessing.image import ImageDataGenerator

20 [22]: # 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)

20 [23]: # Training Dataset
x_train=train_datagen.flow_from_directory("/content/drive/MyDrive/Dataset/training_set",target_size=(64,64), class_mode='categorical',batch_size=66)
# Testing Dataset
x_test=test_datagen.flow_from_directory("/content/drive/MyDrive/Dataset/test_set",target_size=(64,64), class_mode='categorical',batch_size=66)

Found 15768 images belonging to 5 classes.
Found 1296 images belonging to 5 classes.

20 [24]: print("len x-train : ", len(x_train))
print("len x-test : ", len(x_test))

len x-train : 18
len x-test : 3

20 [25]: # The Class Indices in training dataset
x_train.class_indices

Out[25]: {'A': 0, 'B': 1, 'C': 2, 'D': 3, 'E': 4, 'F': 5, 'H': 6, 'M': 7, 'X': 8}
```

Model Creation

```
20 [26]: # Importing libraries
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Convolution2D,MaxPooling2D,Flatten,Dense

20 [27]: # Creating Model
model=Sequential()

20 [28]: # adding layers
model.add(Convolution2D(11,(3,3),activation='relu',input_shape=(64,64,3)))

20 [29]: # Adding layers
model.add(Convolution2D(11,(3,3),activation='relu',input_shape=(64,64,3)))

20 [30]: model.add(MaxPooling2D(pool_size=(2,2)))

20 [31]: model.add(Flatten())

20 [32]: # Adding Dense Layers
model.add(Dense(100,activation='relu'))
model.add(Dense(100,activation='relu'))
model.add(Dense(5,activation='softmax'))

20 [33]: # Compiling the Model
model.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['accuracy'])

20 [34]: # Fitting the Model Generator
model.fit_generator(x_train,steps_per_epoch=len(x_train),epochs=10,validation_data=x_test,validation_steps=len(x_test))

/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
18/18 [=====] - 57s 5s/step - loss: 0.0054 - accuracy: 0.9991 - val_loss: 0.3706 - val_accuracy: 0.9794
Epoch 2/10
18/18 [=====] - 57s 5s/step - loss: 0.0039 - accuracy: 0.9996 - val_loss: 0.1347 - val_accuracy: 0.9791
Epoch 3/10
18/18 [=====] - 55s 5s/step - loss: 0.0036 - accuracy: 0.9996 - val_loss: 0.1128 - val_accuracy: 0.9756
Epoch 4/10
18/18 [=====] - 56s 3s/step - loss: 0.0033 - accuracy: 0.9996 - val_loss: 0.3712 - val_accuracy: 0.9787
Epoch 5/10
18/18 [=====] - 55s 5s/step - loss: 0.0033 - accuracy: 0.9995 - val_loss: 0.3811 - val_accuracy: 0.9784
Epoch 6/10
18/18 [=====] - 55s 5s/step - loss: 0.0029 - accuracy: 0.9997 - val_loss: 0.2758 - val_accuracy: 0.9768
Epoch 7/10
18/18 [=====] - 56s 5s/step - loss: 0.0028 - accuracy: 0.9997 - val_loss: 0.3056 - val_accuracy: 0.9768
Epoch 8/10
18/18 [=====] - 55s 5s/step - loss: 0.0021 - accuracy: 0.9997 - val_loss: 0.3112 - val_accuracy: 0.9768
Epoch 9/10
18/18 [=====] - 55s 5s/step - loss: 0.0020 - accuracy: 0.9997 - val_loss: 0.3236 - val_accuracy: 0.9768
Epoch 10/10
18/18 [=====] - 55s 5s/step - loss: 0.0019 - accuracy: 0.9997 - val_loss: 0.3428 - val_accuracy: 0.9768

Out[34]:
```

```

Saving the Model


In [13]: model.save('x1_model_64_64.h5')

Testing the model

In [14]: import numpy as np
from tensorflow.keras.models import load_model
from tensorflow.keras.preprocessing import image

In [14]: model=load_model('x1_model_64_64.h5')
img=image.load_img('/content/drive/MyDrive/Dataset/test_set/0/2.png',
                  target_size=(64,64))

In [15]: img

Out[15]: 

In [16]: x=image.img_to_array(img)

In [17]: x.ndim

Out[17]: 4

In [18]: x=np.expand_dims(x,axis=0)

In [19]: x.ndim

Out[19]: 5

In [20]: pred=np.argmax(model.predict(x),axis=1)

1/1 [=====] - 0s 140ms/step

In [20]: pred

Out[20]: array([0])

In [20]: index=['A','B','C','D','E','F','G','H','I']
print(index[pred[0]])

D

OPEN CV

In [21]: import cv2


In [22]: img=cv2.imread('/content/drive/MyDrive/Dataset/test_set/0/2.png',1)

In [23]: img2=cv2.imread('/content/drive/MyDrive/Dataset/test_set/0/2.png',0)

In [24]: print(img.shape)

(64, 64, 3)

In [25]: from google.colab.patches import cv2_imshow
cv2_imshow(img)
cv2.waitKey(0)
cv2.destroyAllWindows()



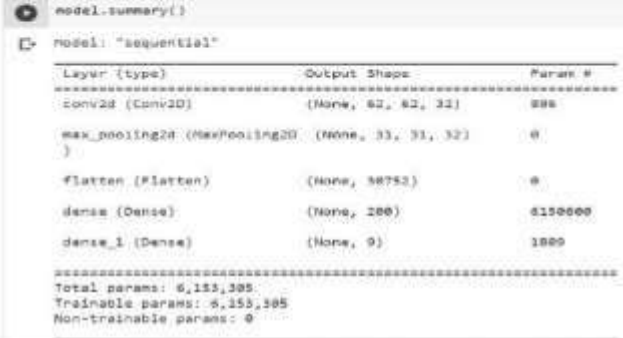
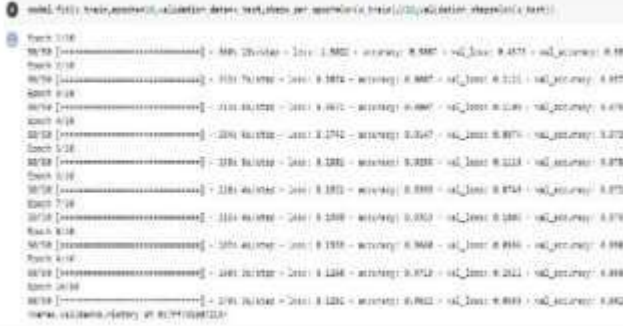
```

9 RESULTS

9.1 Performance metrics

Model Performance Testing:

Project team shall fill the following information in model performance testing template.

S.No.	Parameter	Values	Screenshot
1.	Model Summary	Model - Sequential model Layers: Conv2D-(None,62,62,32) MaxPooling2D-(None,31,31,32) Flatten-(None,30752) Dense-(None,200) Dense_1-(None,9)	 <pre>model.summary() model: "sequential" Layer (type) Output Shape Param # ----- conv2d (Conv2D) (None, 62, 62, 32) 888 max_pooling2d (MaxPooling2D) (None, 31, 31, 32) 0 flatten (Flatten) (None, 98752) 0 dense (Dense) (None, 200) 6150000 dense_1 (Dense) (None, 9) 1800 ----- Total params: 6,155,305 Trainable params: 6,155,305 Non-trainable params: 0</pre>
2.	Accuracy	Training Accuracy - 0.9622 Validation Accuracy -0.9826	 <pre>model.fit(x=train_generator(), validation_data=test_data, epochs=10, callbacks=[callbacks], verbose=1) Epoch 1/10 10/10 [=====] - loss: 1.8602 - accuracy: 0.9607 - val_loss: 0.4575 - val_accuracy: 0.9812 Epoch 2/10 10/10 [=====] - loss: 0.3876 - accuracy: 0.9867 - val_loss: 0.3215 - val_accuracy: 0.9875 Epoch 3/10 10/10 [=====] - loss: 0.4070 - accuracy: 0.9867 - val_loss: 0.3240 - val_accuracy: 0.9876 Epoch 4/10 10/10 [=====] - loss: 0.4070 - accuracy: 0.9867 - val_loss: 0.3240 - val_accuracy: 0.9876 Epoch 5/10 10/10 [=====] - loss: 0.3242 - accuracy: 0.9867 - val_loss: 0.3240 - val_accuracy: 0.9876 Epoch 6/10 10/10 [=====] - loss: 0.3242 - accuracy: 0.9867 - val_loss: 0.3240 - val_accuracy: 0.9876 Epoch 7/10 10/10 [=====] - loss: 0.3242 - accuracy: 0.9867 - val_loss: 0.3240 - val_accuracy: 0.9876 Epoch 8/10 10/10 [=====] - loss: 0.3242 - accuracy: 0.9867 - val_loss: 0.3240 - val_accuracy: 0.9876 Epoch 9/10 10/10 [=====] - loss: 0.3242 - accuracy: 0.9867 - val_loss: 0.3240 - val_accuracy: 0.9876 Epoch 10/10 10/10 [=====] - loss: 0.3242 - accuracy: 0.9867 - val_loss: 0.3240 - val_accuracy: 0.9876 Total 10 epochs training complete. Training loss: 0.3242, Validation loss: 0.3240, Training accuracy: 0.9867, Validation accuracy: 0.9876</pre>
3	Confidence Score	Class Detected - N/A Confidence Score -N/A	N/A

10. ADVANTAGES AND DISADVANTAGES

ADVANTAGES

It enables employees from across the world to communicate with each other 24x7 and share ideas or solve problems quickly. It is a cost-effective way of getting several people from different locations to attend meetings and conferences—without having to spend time or money on travel and accommodation.

DISADVANTAGES

The biggest disadvantage of communication is that it takes a lot of time to listen, speak, read, or write to someone. While trying to do one thing, you can accidentally hurt another person's feelings by not listening or paying attention. This could result in damaging your relationship with them.

11. CONCLUSION

Real-time communication (RTC) workloads can be deployed on AWS to attain scalability, elasticity, and high availability while meeting the key requirements. Today, several customers are using AWS, its partners, and open source solutions to run RTC workloads with reduced cost and faster agility as well as a reduced global footprint. The reference architectures and best practices provided in this white paper can help customers successfully set up RTC workloads on AWS and optimize the solutions to meet end-user requirements while optimizing for the cloud.

12. FUTURE SCOPE

1. Through image recognition technology, AI understands the context of objects in photos and describes photos to people.

2. The speech-to-text and text-to-speech technologies helped those people who had speech impediments.

3. The product in AI that narrates the entire world around them visually impaired by reading texts describing whereabouts and the looks of the nearby people by identifying and recognizing faces and emotions.

4. Autonomous vehicles are in trend and their success is due to AI technology. These vehicles can be beneficial to people living with limited physical mobility.

13. APPENDIX

Source code

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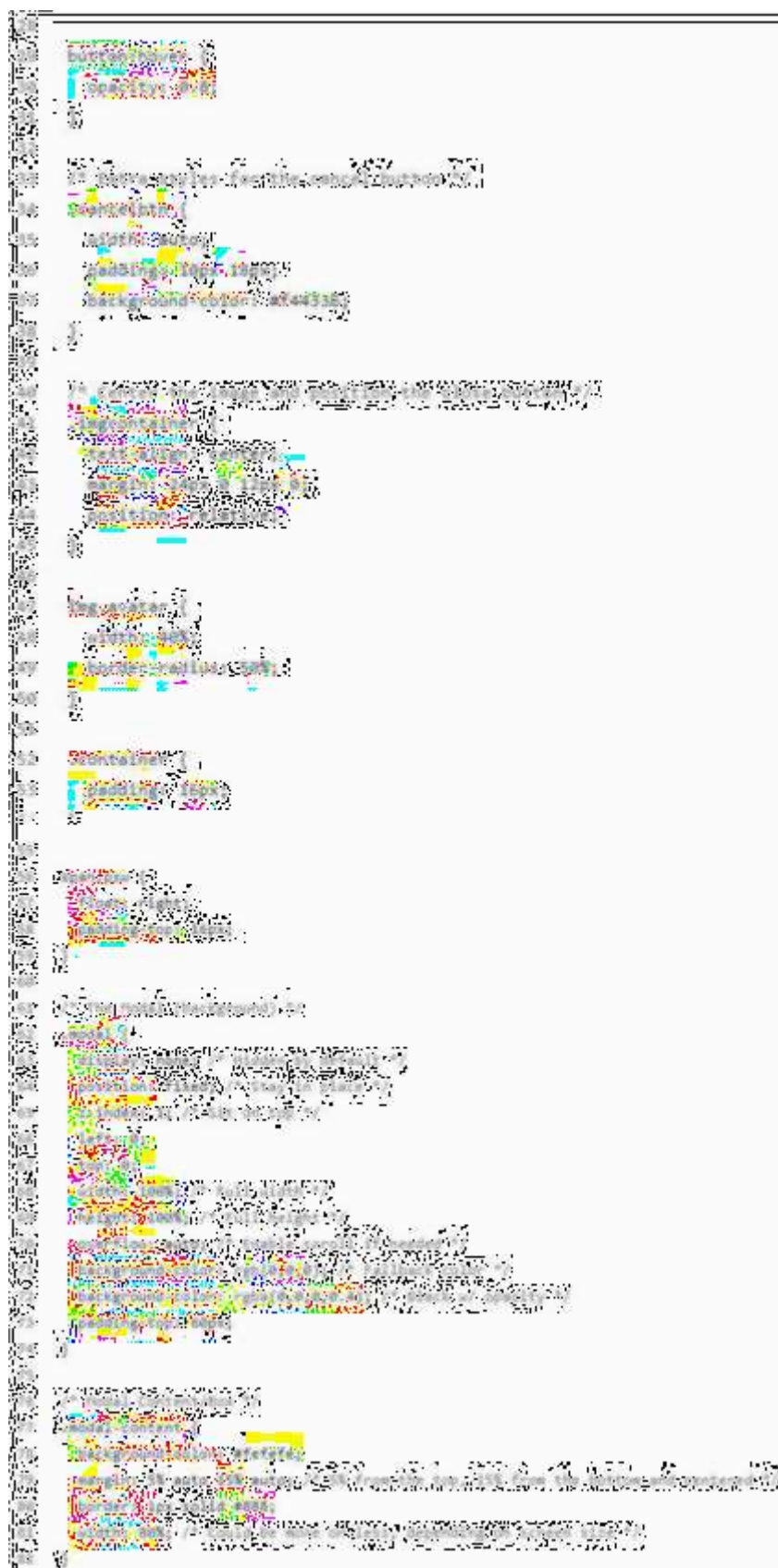
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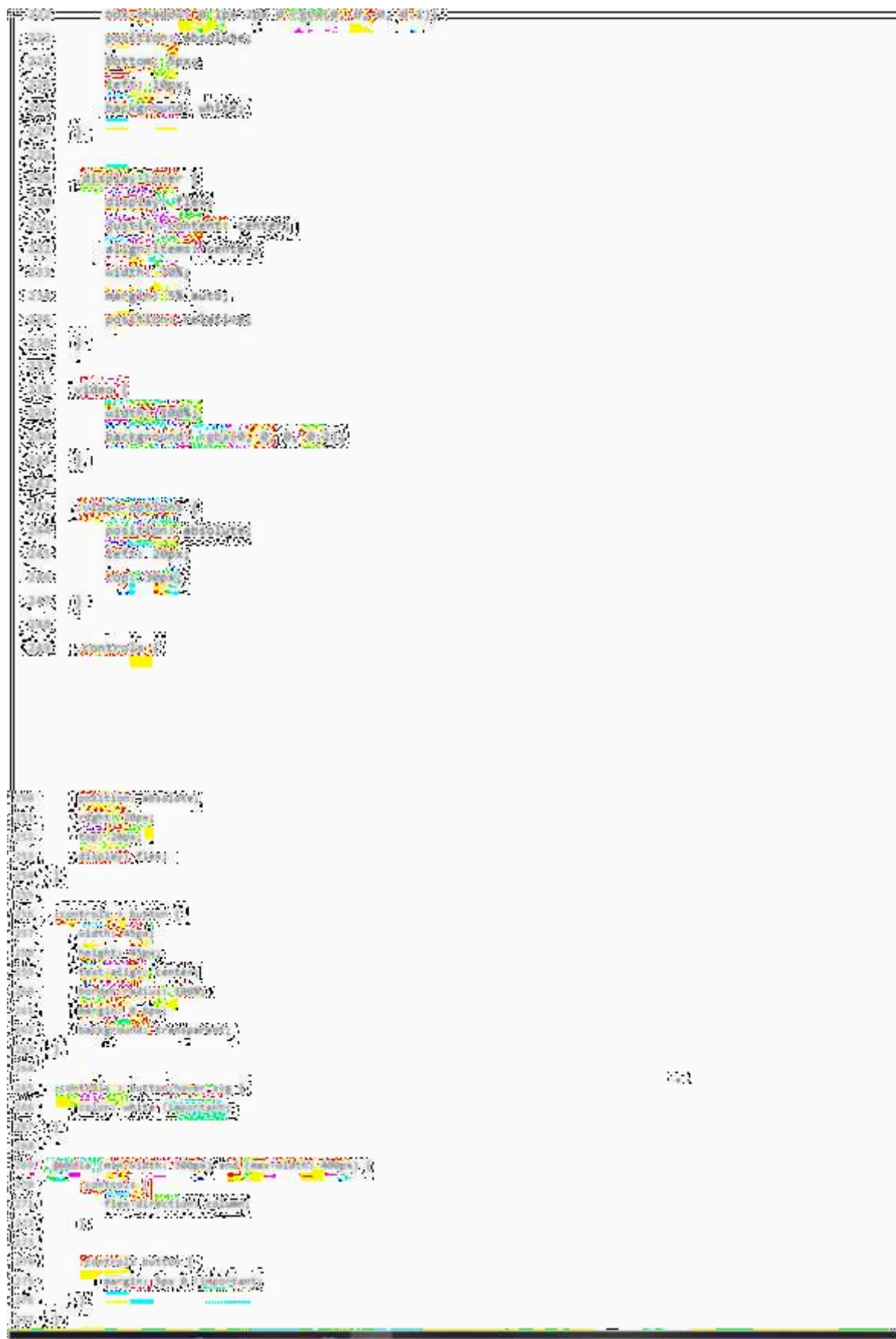
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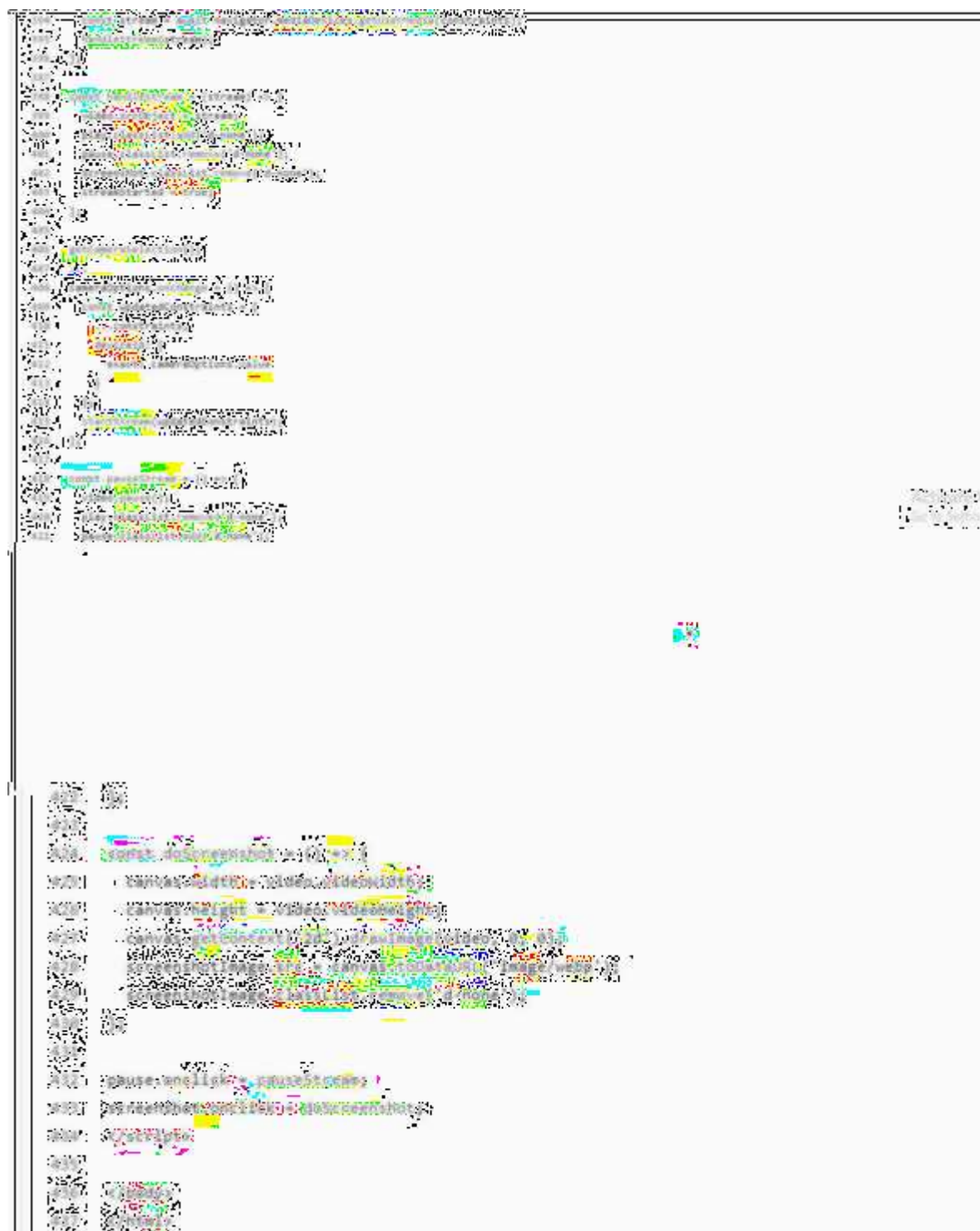
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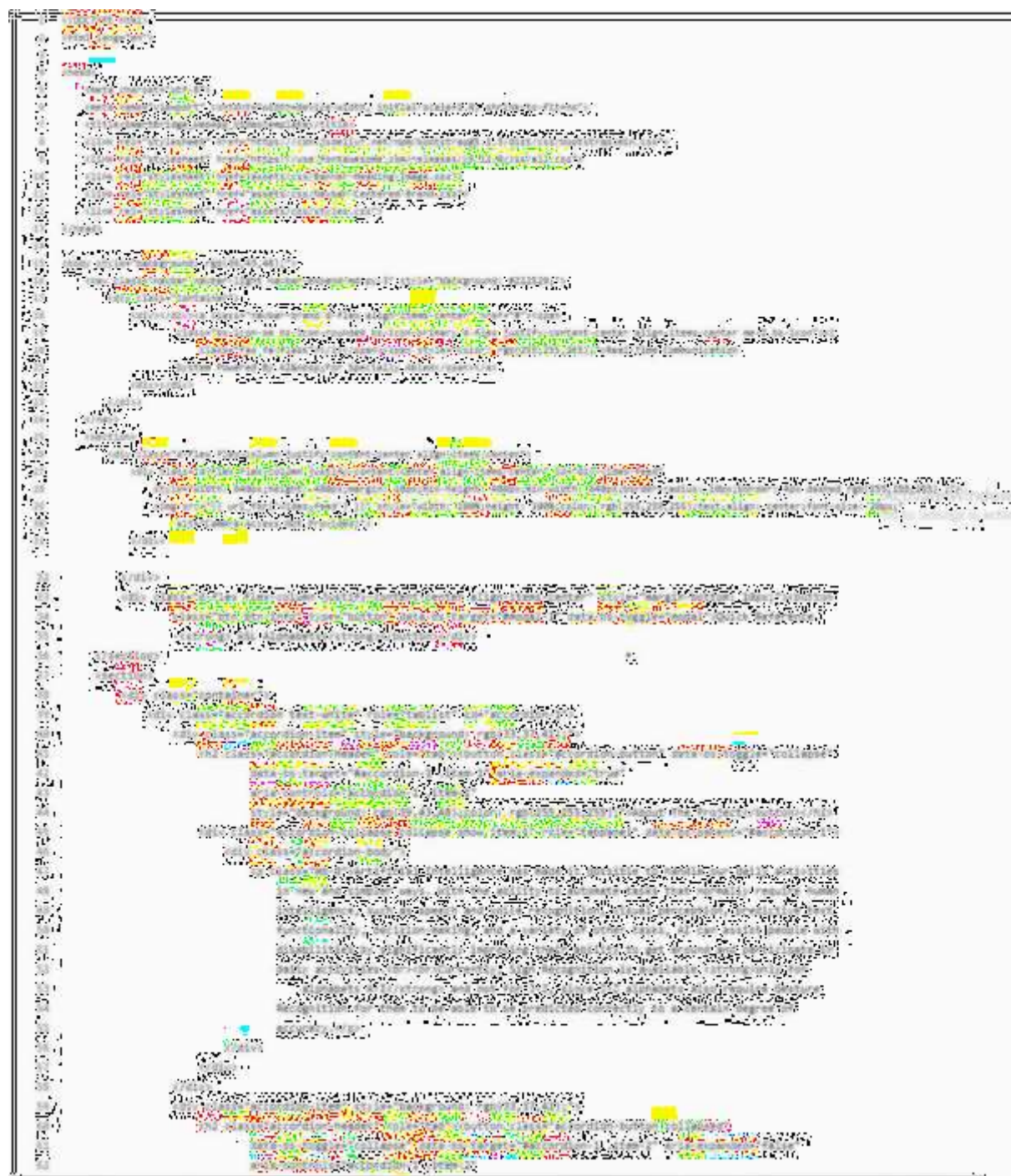


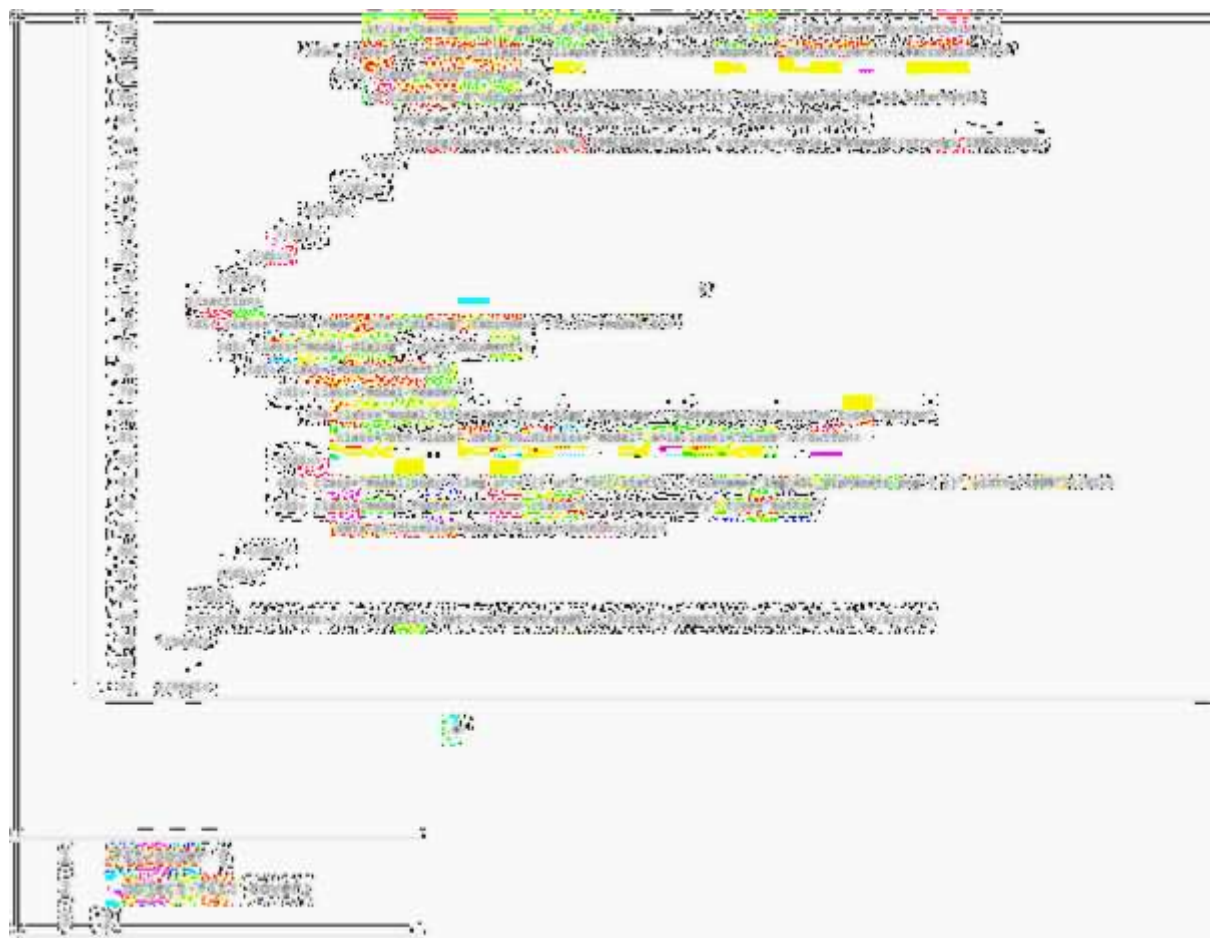
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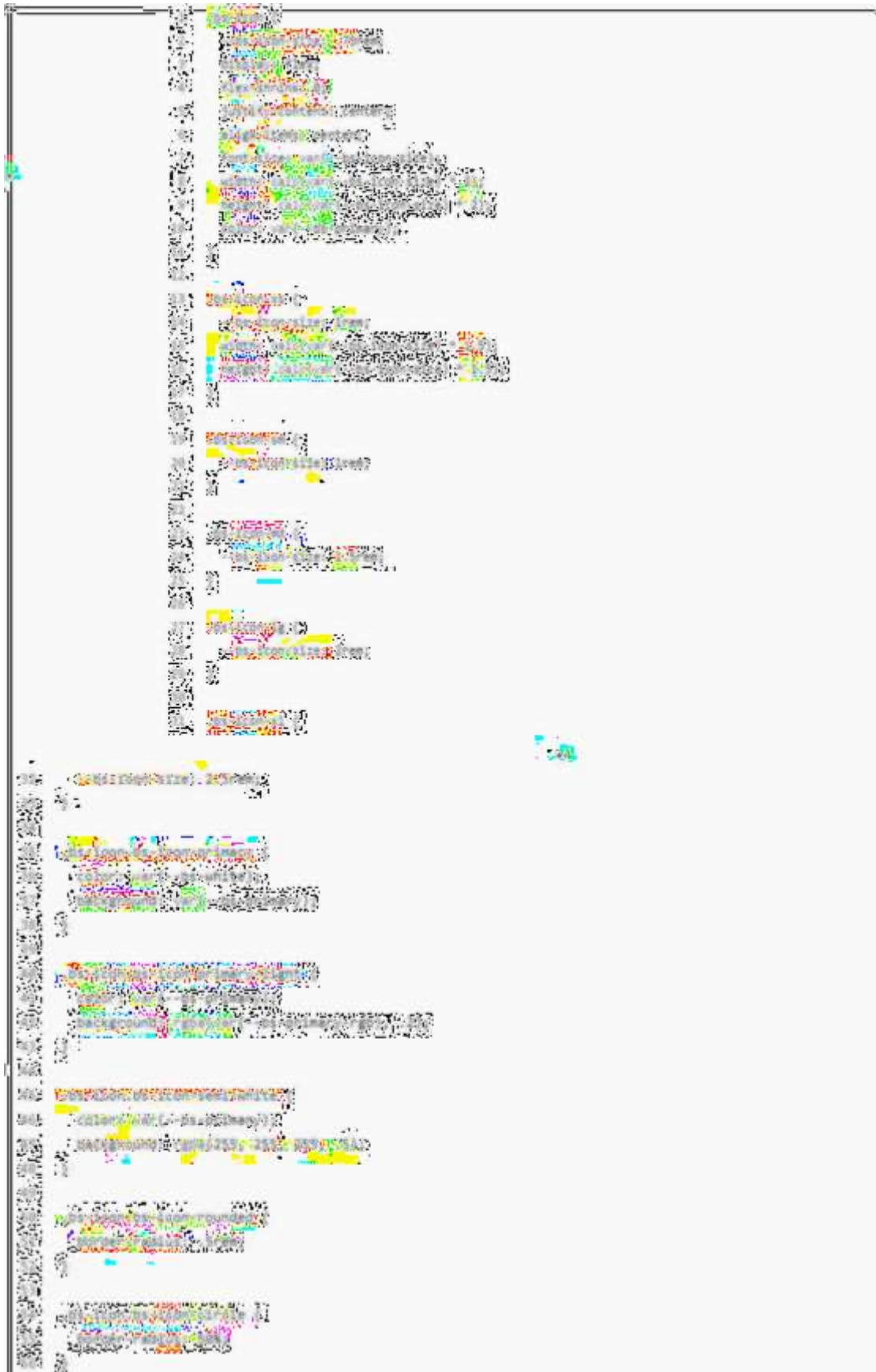


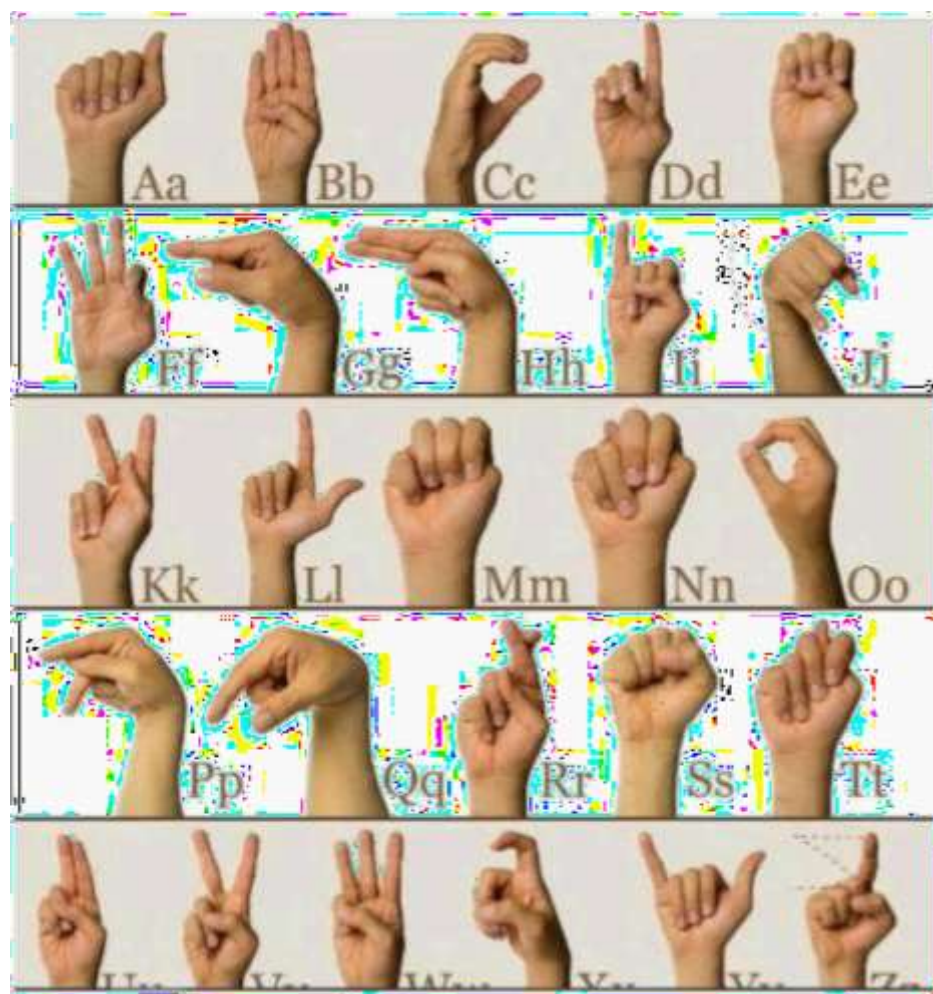














PROJECT DEMO LINK

<https://drive.google.com/drive/folders/1X7M2if1rQura9N6CrWdzT0rPOM-c4yBC>

GITHUB LINK

<https://github.com/IBM-EPBL/IBM-Project-8147-1658910120>