# PROJECT REPORT

# A NOVEL METHOD FOR HANDWRITTEN DIGIT RECOGNITION

# submitted by PNT2022TMID03378

Sudarson A.M - 212219040154

Sanjay Kumar Y - 212219040135

Sumer Singh B - 212219040158

Praveen Kumar G - 212219040112

# **TABLE OF CONTENTS**

1 INTRODUCTION	1
1.1 PROJECT OVERVIEW	1
1.2 PURPOSE	1
2 LITERATURE SURVEY	2
2.1 EXISTING PROBLEM	2
2.2 REFERENCES	2
23 PROBLEM STATEMENT DEFINITION	5
3 IDEATION AND PROPOSED SOLUTION	6
3.1 EMPATHY MAP CANVAS	6
3.2 IDEATION & BRAINSTORMING	7
3.3 PROPOSED SOLUTION	8
3.4 PROBLEM SOLUTION FIT	9
4 REQUIREMENT ANALYSIS	10
4.1 FUNCTIONAL REQUIREMENTS	10
4.2 NON FUNCTIONAL REQUIREMENTS	11
5 PROJECT DESIGN	12
5.1 DATA FLOW DIAGRAM	12
5.2 SOLUTION & TECHNICAL ARCHITECTURE	13
5.3 USER STORIES	15
6 PROJECT PLANNING AND SCHEDULING	16

6.1 SPRINT PLANNING AND ESTIMATION	16
6.2 SPRINT DELIVERY SCHEDULE	17
7 CODING & SOLUTIONING	18
8 TESTING	20
8.1 TEST CASES	20
8.2 USER ACCEPTANCE TESTING	22
82.1 DEFECT ANALYSIS	22
82.2 TEST CASE ANALYSIS	22
9 RESULTS	23
9.1 PERFORMANCE METRICS	23
10 ADVANTAGES & DISADVANTAGES	25
ADVANTAGES	25
DISADVANTAGES	25
11 CONCLUSION	26
12 FUTURE SCOPE	27
APPENDIX	28
SOURCE CODE	28
GITHUB	37
PROJECT DEMO	37

# CHAPTER 1 INTRODUCTION

### 1.1 PROJECT OVERVIEW

Machine learning and deep education play an main duty in the study of computers and artificial intelligence. With the use of deep education and machine intelligence, human exertion maybe diminished in identifying, learning, forecastings and in many more extents.

Handwritten Digit Recognition is the capability of calculating plans to recognise in manuscript digits from differing sources, in the way that figures, documents, thus. This project aims unoccupied consumers impose upon machine intelligence to reduce manual tasks in seeing digits.

### 1.2 PURPOSE

Digit acknowledgment systems are worthy making the digits from different beginnings like emails, bank check for paying money, papers, countenances, etc. and indifferent real-realm sketches for online script acknowledgment on computer tablets or method, understand; number plates of vehicles, deal with bank check for paying money amounts, numeric efforts in forms filled up manually (tax forms) thus.

# CHAPTER 2 LITERATURE SURVEY

### 21 EXISTING PROBLEM

The fundamental problem with handwritten digit recognition is that handwritten digits do not always have the same size, width, orientation, and marginssince they vary from person to person. Additionally, there would be issues with identifying the numbers because of similarities between numerals like 1 and 7, 5 and 6, 3 and 8, 2 and 5, 2 and 7, etc. Finally, the individuality and variation of each individual's handwriting influence the structure and appearance of the digits.

### 2.2 REFERENCES

Improved Handwritten Digit Recognition Using Convolutional Neural Networks(CNN) (2020)

Ahlawat, Savita and Choudhary, Amit and Nayyar, Anand and Singh, Saurabh and Yoon, Byungun

This paper's primary goal was to enhance handwritten digit recognition ability. To avoid difficult pre-processing, expensive feature extraction, and a complex ensemble (classifier combination) method of a standard recognition system, they examined different convolutional neural network variations. Their current work makes suggestions on the function of several hyper-parameters through thoroughevaluation utilizing an MNIST dataset. They also confirmed that optimizing hyper-parameters is crucial for enhancing CNN architecture performance. With the Adam optimizer for the MNIST database, they were able to surpass many previously published results with a recognition rate of 99.89%. Through the trials, it is made

abundantly evident how the performance of handwritten digit recognition is affected by the number of convolutional layers in CNN architecture. According to the paper, evolutionary algorithms can be explored for optimizing convolutional filterkernel sizes, CNN learning parameters, and the quantity of layers and learning rates.

# An Efficient And Improved Scheme For Handwritten Digit Recognition Based On Convolutional Neural Network (2019)

Ali, Saqib and Shaukat, Zeeshan and Azeem, Muhammad and Sakhawat, Zareen and Mahmood, Tariq and others

This study uses rectified linear units (ReLU) activation and a convolutional neural network (CNN) that incorporates the Deeplearning4j (DL4J) architecture to recognize handwritten digits. The proposed CNN framework has all the necessary parameters for a high level of MNIST digit classification accuracy. The system's training takes into account the time factor as well. The system is also tested by altering the number of CNN layers for additional accuracy verification. It is important to note that the CNN architecture consists of two convolutional layers, the first with 32 filters and a 5x5 window size and the second with 64 filters and a 7x7 window size. In comparison to earlier proposed systems, the experimental findings show that the proposed CNN architecture for the MNIST dataset demonstrates great performance in terms of time and accuracy. As a result, handwritten numbers are detected with a recognition rate of 99.89% and high precision (99.21%) in a short amount of time.

# Improved Handwritten Digit Recognition Using Quantum K-Nearest Neighbor Algorithm (2019)

Wang, Yuxiang and Wang, Ruijin and Li, Dongfen and Adu-Gyamfi, Daniel and Tian, Kaibin and Zhu, Yixin

The KNN classical machine learning technique is used in this research to enable quantum parallel computing and superposition. They used the KNN algorithm with quantum acceleration to enhance handwritten digit recognition. When dealing with more complicated and sizable handwritten digital data sets, their suggested methodconsiderably lowered the computational time complexity of the traditional KNN algorithm. The paper offered a theoretical investigation of how quantum concepts can be applied to machine learning. Finally, they established a fundamental operational concept and procedure for machine learning with quantum acceleration.

### Handwritten Digit Recognition Using Machine And Deep Learning Algorithms (2021)

Pashine, Samay and Dixit, Ritik and Kushwah, Rishika

In this study, they developed three deep and machine learning-based models for handwritten digit recognition using MNIST datasets. To determine which model wasthe most accurate, they compared them based on their individual properties.

Support vector machines are among the simplest classifiers, making them faster than other algorithms and providing the highest training accuracy rate in this situation. However, due to their simplicity, SVMs cannot categorize complicated and ambiguous images as accurately as MLP and CNN algorithms can. In their research, they discovered that CNN produced the most precise outcomes for handwritten digit recognition. This led them to the conclusion that CNN is the most effective

solution for all types of prediction issues, including those using picture data. Next, by comparing the execution times of the algorithms, they determined that increasing the number of epochs without changing the configuration of the algorithm is pointless due to the limitation of a certain model, and they discoveredthat beyond a certain number of epochs, the model begins over-fitting the dataset and provides biased predictions.

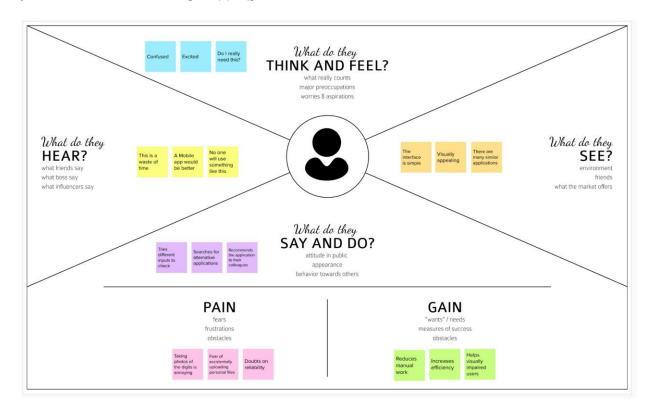
### 23 PROBLEM STATEMENT DEFINITION

For years, the traffic department has been combating traffic law violators.

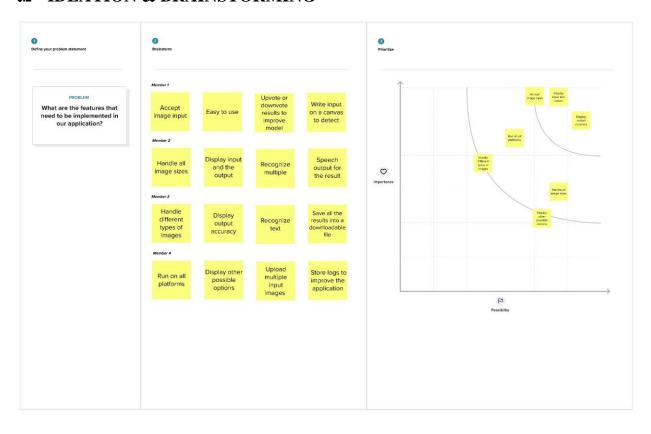
These offenders endanger not only their own lives, but also the lives of other individuals. Punishing these offenders is critical to ensuring that others do not become like them. Identification of these offenders is next to impossible because it is impossible for the average individual to write down the license plate of a recklessdriver. Therefore, the goal of this project is to help the traffic department identify these offenders and reduce traffic violations as a result.

# CHAPTER 3 IDEATION AND PROPOSED SOLUTION

# 3.1 EMPATHY MAP CANVAS



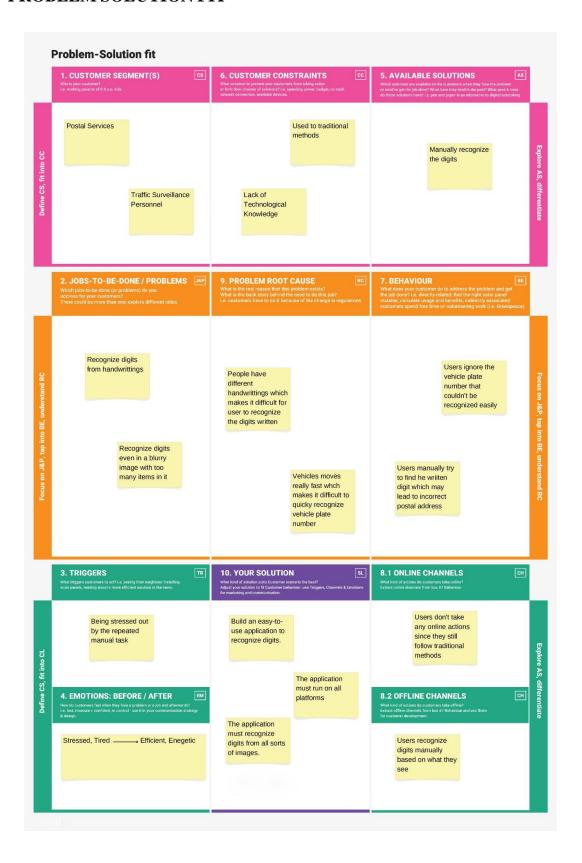
# 3.2 IDEATION & BRAINSTORMING



# 3.3 PROPOSED SOLUTION

S.NO	PARAMETER	DESCRIPTION
1	Problem Statement	To create an application that recognizes handwritten digits
2	Idea / Solution Description	The application takes an imageas the input and accurately detects the digits in it.
3	Novelty/Uniqueness	Instead of recognizing every text, the application accurately recognizes only the digits
4	Social Impact / Customer Satisfaction	This application reduces the manual tasks that need to be performed. This improves productivity in the workplace.
5	Business Model	The application can be integrated with traffic surveillance cameras to recognize vehicle number plates  The application can be integrated with Postal systemsto recognize the pin codes effectively
6	Scalability of the Solution	The application can easily be scaled to accept multiple inputs and process them parallelly to further increase efficiency

### 3.4 PROBLEM SOLUTION FIT



# CHAPTER 4 REQUIREMENT ANALYSIS

# 4.1 FUNCTIONAL REQUIREMENTS

FR.NO	FUNCTIONAL REQUIREMENTS	SUB REQUIREMENTS
		Get access the MNIST dataset
FR-1	Model Creation	Analyze the dataset
rk-i	Woder Creation	Define a CNN model
		Train and Test the Model
		Create a website to let the user recognize handwritten digits.
	Application Development	Create a home page to upload images
FR-2		Create a result page to displaythe results
		Host the website to let the usersuse it from anywhere
		Let users upload images of various formats.
FR-3		Let users upload images of various size
	Input Image Upload	Prevent users from uploading unsupported image formats
		Pre-Process the image to use iton the model

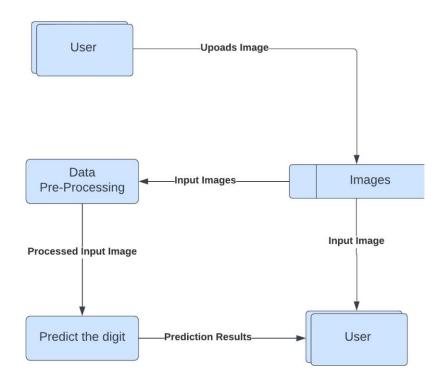
		Create a database to store allthe input images	
		Display the result from themodel	
		Display input image	
FR-4	Display Results	Display accuracy the result	
		Display other possible predictions with their respective accuracy	

# 4.2 NON FUNCTIONAL REQUIREMENTS

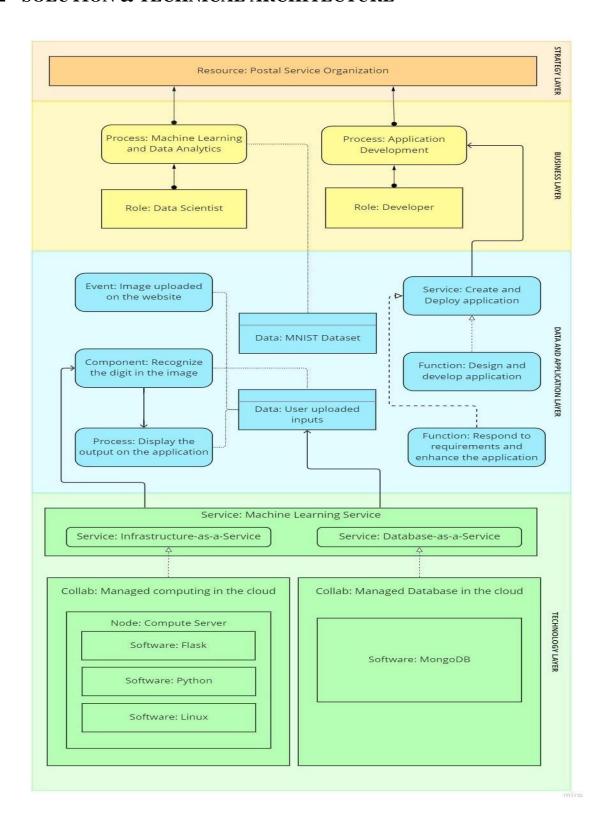
NFR	NON-FUNCTIONAL REQUIREMENTS	DESCRIPTION
NFR-l	Usability	The application must be usablein all devices
NFR-2	Security	The application must protectuser uploaded image
NFR-3	Reliability	The application must give anaccurate result as much as possible
NFR-4	Performance	The application must be fast andquick to load up
NFR-5	Availability	The application must be available to use all the time
NFR-6	Scalability	The application must scale alongwith the user base

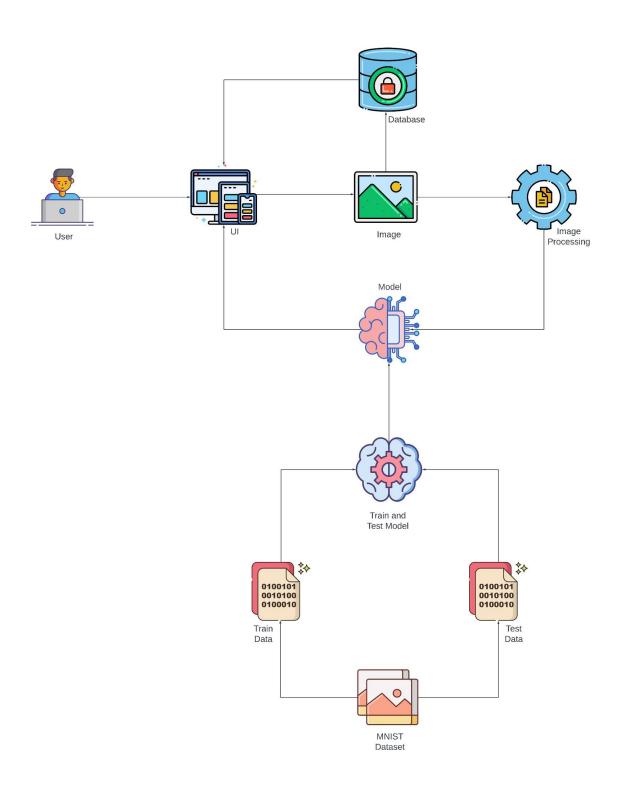
# CHAPTER 5 PROJECT DESIGN

# **5.1 DATA FLOW DIAGRAM**



# 5.2 SOLUTION & TECHNICAL ARCHITECTURE





# 5.3 USER STORIES

User Type	Functional Requirements	User Story Number	User Story / Task	Acceptance Criteria	Priority	Release
	Accessing the Application	USN-I	As a user, I should beable to access the application from anywhere and use onany devices	User can access the application using the browser on any device	High	Sprint-4
	Uploading Image	USN-2	As a user, I should be able to upload images to predict the digits	User can upload images	High	Sprint-3
Customer	Viewing the Results	USN-3	As a user, I should beable to view the results	The result of the prediction is displayed	High	Sprint-3
	Viewing Other Prediction	USN-4	As a user, I should beable to see other close predictions	The accuracy of other values must be displayed	Medium	Sprint-4
	Usage Instruction	USN-5	As a user, I should have a usage instruction to know how to use the application	The usage instruction is displayed on the home page	Medium	Sprint-4

# CHAPTER 6 PROJECT PLANNING AND SCHEDULING

# **6.1 SPRINT PLANNING AND ESTIMATION**

SPRINT	USER STORY / TASK	STORY POINTS	PRIORITY	TEAM MEMBERS
	Get the dataset	3	High	Sudarson A.M
	Explore the data	2	Medium	Sudarson A.M
	Data Pre-Processing	3	High	Sudarson A.M
Sprint - I	Prepare training and testing data	3	High	Sudarson A.M
	Create the model	3	High	Sanjay Kumar Y
	Train the model	3	High	Sanjay Kumar Y
Sprint - II	Test the model	3	High	Sanjay Kumar Y
	Improve the model	2	Medium	Sumer Singh B
	Save the model	3	High	Sumer Singh B
	Build the Home Page	3	High	Sumer Singh B
Sprint - III Setup a database tostore input images		2	Medium	Sumer Singh B
Sprint - IV	Build the results page	3	High	Praveen Kumar G

Integrate the modelwith the application	3	High	Praveen Kumar G
Test the application	3	High	Praveen Kumar G

# 6.2 SPRINT DELIVERY SCHEDULE

SPRINT	TOTA L STORY POINTS	DURATIO N	SPRINT START DATE	SPRINT END DATE (PLANNED)	STORY POINTS COMPLETE D(AS ON PLANNED DATE)	SPRINT RELEASE DATE (ACTUAL)
Sprint - I	20	6 Days	24 Oct 2022	29 Oct 2022	20	29 Oct 2022
Sprint - II	20	6 Days	31 Oct 2022	05 Nov 2022	20	05 Nov 2022
Sprint - III	20	6 Days	07 Oct 2022	12 Nov 2022	20	12 Nov 2022
Sprint - IV	20	6 Days	14 Nov 2022	19 Nov 2022	20	19 Nov 2022

# CHAPTER 7 CODING & SOLUTIONING

```
# Import necessary packages
import os
import random
import string
from pathlib import Path
import numpy as np
from tensorflow.keras.models import load_model
from PIL import Image, ImageOps
```

```
def random_name_generator(n: int) -> str:
    """
    Generates a random file name.

Args:
    n (int): Length the of the file name.

Returns:
    str: The file name.
"""
    return ''.join(random.choices(string.ascii_uppercase + string.digits, k=n))
```

```
def recognize(image: bytes) -> tuple:
   model=load model(Path("./model/model.h5"))
   img = Image.open(image).convert("L")
   img_name = random_name_generator(10) + '.jpg'
   img.save(Path(f"./static/data/{img_name}"))
   img = ImageOps.grayscale(img)
    img = ImageOps.invert(img)
   img = img.resize((28, 28))
   img2arr = np.array(img)
   img2arr = img2arr / 255.0
   img2arr = img2arr.reshape(1, 28, 28, 1)
   results = model.predict(img2arr)
   best = np.argmax(results,axis = 1)[0]
   pred = list(map(lambda x: round(x*100, 2), results[0]))
   best = others.pop(best)
   return best, others, img_name
```

# CHAPTER 8 TESTING

# 8.1 TEST CASES

Test case ID	Feature Type	Component	Test Scenario	Expected Result	Actual Result	Status
HP_TC_00l	UI	Home Page	Verify UI elements in the Home Page	The Home page must be displayedproperly	Working as expected	PASS
HP_TC_002	UI	Home Page	Check if the UI elements are displayed properly in different screen sizes	The Home page must be displayed properly in all sizes	The UI is not displayed properly in screen size 2560 x  80  and 768 x 630	FAIL
HP_TC_003	Functional	Home Page	Check if usercan upload their file	The input imageshould be uploaded to the application successfully	Working as expected	PASS
HP_TC_004	Functional	Home Page	Check if user cannot upload unsupported files	The application should not allow user to select a non image file	User is able to upload any file	FAIL
HP_TC_005	Functional	Home Page	Check if the page redirectsto the result page once theinput is given	The page should redirect to the results page	Working as expected	PASS

BE_TC_001	Functional	Backend	Check if all the routes are working properly	All the routes should properlywork	Working as expected	PASS
M_TC_001	Functional	Model	Check if the model can handle various image sizes	The model should rescale the imageand predict the results	Working as expected	PASS
M_TC_002	Functional	Model	Check if the model predicts thedigit	The model should predict the number	Working as expected	PASS
M_TC_003	Functional	Model	Check if the model can handle complex inputimage	The model should predict the number in the complex image	The model failsto identify the digit since the model is not built to handle such data	FAIL
RP_TC_001	UI	Result Page	Verify UI elements in the Result Page	The Result page must be displayed properly	Working as expected	PASS
RP_TC_002	UI	Result Page	Check if the input image is displayed properly	The input image should be displayed properly	The size of the input image exceeds the display container	FAIL
RP_TC_003	UI	Result Page	Check if theresult is displayed properly	The result shouldbe displayed properly	Working as expected	PASS
RP_TC_004	UI	Result Page	Check if the other predictions are displayed properly	The other predictions shouldbe displayed properly	Working as expected	PASS

# 8.2 USER ACCEPTANCE TESTING

# 8.2.1 DEFECT ANALYSIS

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Total
By Design	1	0	1	0	2
Duplicate	0	0	0	0	0
External	0	0	2	0	2
Fixed	4	1	0	1	6
Not Reproduced	0	0	0	1	1
Skipped	0	0	0	1	1
Won't Fix	1	0	1	0	2
Total	6	1	4	3	14

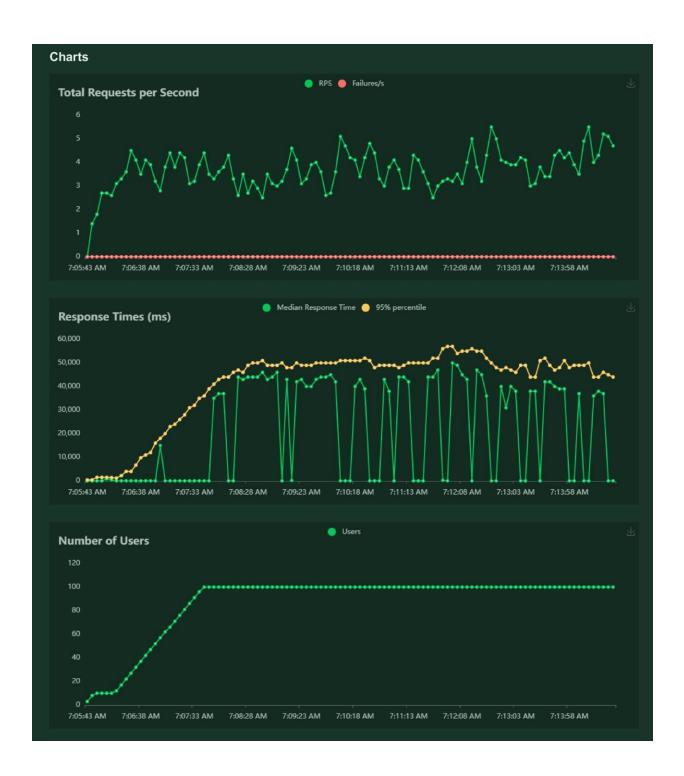
# 8.22 TEST CASE ANALYSIS

Section	Total Cases	Not Tested	Fail	Pass
Client Application	10	0	3	7
Security	2	0	1	1
Performance	3	0	1	2
Exception Reporting	2	0	0	2

# CHAPTER 9 RESULTS

# 9.1 PERFORMANCE METRICS





# CHAPTER 10 ADVANTAGES & DISADVANTAGES

### **ADVANTAGES**

- Reduces manual work
- More accurate than average human
- Capable of handling a lot of data
- Can be used anywhere from any device

# **DISADVANTAGES**

- Cannot handle complex data
- All the data must be in digital format
- Requires a high performance server for faster predictions
- Prone to occasional errors

# CHAPTER 11 CONCLUSION

This project demonstrated a web application that uses machine learning to recognise handwritten numbers. Flask, HTML, CSS, JavaScript, and a few other technologies were used to create this project. The model predicts the handwritten digit using a CNN network. During testing, the model achieved a 99.61% recognition rate. The proposed project is scalable and can easily handle a huge number of users. Since it is a web application, it is compatible with any device that can run a browser. This project is extremely useful in real-world scenarios such as recognizing numberplates of vehicles, processing bank cheque amounts, numeric entries in forms filled up by hand (tax forms) and so on. There is so much room for improvement, which can be implemented in subsequent versions.

# CHAPTER 12 FUTURE SCOPE

This project is far from complete and there is a lot of room for improvement.

Some of the improvements that can be made to this project are as follows:

- Add support to detect from digits multiple images and save the results
- Add support to detect multiple digits
- Improve model to detect digits from complex images
- Add support to different languages to help users from all over the world

This project has endless potential and can always be enhanced to become better.Implementing this concept in the real world will benefit several industries and reduce the workload on many workers, enhancing overall work efficiency.

# **APPENDIX**

### **SOURCE CODE**

#### MODEL CREATION

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from keras.utils import np_utils
from tensorflow.keras.datasets import mnist
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, Dense, Flatten
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.models import load_model
from PIL import Image, ImageOps
(X_train, y_train), (X_test, y_test) = mnist.load_data()
X_train = X_train.reshape(60000, 28, 28, 1).astype('float32')
X_test = X_test.reshape(10000, 28, 28, 1).astype('float32')
number_of_classes = 10
Y_train = np_utils.to_categorical(y_train, number_of_classes)
Y_test = np_utils.to_categorical(y_test, number_of_classes)
```

```
# Create the model
model = Sequential()
model.add(conv2D(64, (3, 3), input_shape=(28, 28, 1), activation="relu"))
model.add(Conv2D(32, (3, 3), activation="relu"))
model.add(Flatten())
model.add(Dense(number_of_classes, activation="softmax"))

model.compile(loss='categorical_crossentropy', optimizer="Adam", metrics=["accuracy"])

# Train the model
model.fit(X_train, Y_train, batch_size=32, epochs=5, validation_data=(X_test,Y_test))

# Evaluate the model
metrics = model.evaluate(X_test, Y_test, verbose=0)
print("Metrics (Test Loss & Test Accuracy): ")
print(metrics)

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

```
# Test the saved model
model=load_model("model.h5")

img = Image.open("sample.png").convert("L")
img = img.resize((28, 28))
img2arr = np.array(img)
img2arr = img2arr.reshape(1, 28, 28, 1)
results = model.predict(img2arr)
results = np.argmax(results,axis = 1)
results = pd.Series(results,name="Label")
print(results)
```

#### FLASK APP

```
from flask import Flask,render_template,request
from recognizer import recognize

app=Flask(__name__)

@app.route('/')
def main():
    return render_template("home.html")

@app.route('/predict',methods=['POST'])
def predict():
    if request.method=='POST':
        image = request.files.get('photo', '')
        best, others, img_name = recognize(image)
        return render_template("predict.html", best=best, others=others, img_name=img_name)

if __name__=="__main__":
    app.run()
```

### **RECOGNIZER**

```
# Import necessary packages
import os
import random
import string
from pathlib import Path
import numpy as np
from tensorflow.keras.models import load_model
from PIL import Image, ImageOps
```

```
def random_name_generator(n: int) -> str:
    """
    Generates a random file name.

Args:
    n (int): Length the of the file name.

Returns:
    str: The file name.
"""
    return ''.join(random.choices(string.ascii_uppercase + string.digits, k=n))
```

```
model=load_model(Path("./model/model.h5"))
img = Image.open(image).convert("L")
img_name = random_name_generator(10) + '.jpg'
img.save(Path(f"./static/data/{img_name}"))
img = ImageOps.grayscale(img)
img = ImageOps.invert(img)
img = img.resize((28, 28))
img2arr = np.array(img)
img2arr = img2arr / 255.0
img2arr = img2arr.reshape(1, 28, 28, 1)
results = model.predict(img2arr)
best = np.argmax(results,axis = 1)[0]
pred = list(map(lambda x: round(x*100, 2), results[0]))
others = list(zip(values, pred))
return best, others, img_name
```

#### **HOME PAGE (HTML)**

```
<meta name="viewport" content="width=device-width, initial-scale=1.0" />
<title>Handwritten Digit Recognition</title>
<link rel="icon" type="image/svg" sizes="32x32" href="{{url_for('static',filename='images/icon.svg')}}"</pre>
<link rel="stylesheet" href="{{url_for('static',filename='css/main.css')}}" />
<script src="https://unpkg.com/feather-icons"></script>
<script defer src="{{url_for('static',filename='js/script.js')}}"></script>
<div class="container">
    <div class="heading">
        <h1 class="heading_main">Handwritten Digit Recognizer</h1>
        <h2 class="heading_sub">Easily analyze and detect handwritten digits</h2>
    <div class="upload-container">
        <div class="form-wrapper">
            <form class="upload" action="/predict" method="post" enctype="multipart/form-data">
                <label id="label" for="upload-image"><i data-feather="file-plus"></i>>Select File</label>
                <input type="file" name="photo" id="upload-image" hidden />
                <button type="submit" id="up_btn"></button>
            <img id="loading" src="{{url_for('static',filename='images/loading.gif')}}">
```

### HOME PAGE (CSS)

```
@import url("https://fonts.googleapis.com/css2?family=Overpass:wght@200;300;400;500;600;700;900&display=swap");

* {
    padding: 0;
    margin: 0;
}

body {
    color: black;
    font-family: "Overpass", sans-serif;
}
```

```
.container {
   width: 100%;
   height: 100%;
   flex-direction: column;
.heading {
   margin-top: -2rem;
   padding-bottom: 2rem;
.heading .heading__main {
   font-size: 3rem;
.heading .heading_sub {
   font-size: 1rem;
   color: rgb(90, 88, 88);
.upLoad-container {
   box-shadow: 0 0 20px rgb(172, 170, 170);
   width: 40rem;
   height: 25rem;
   padding: 1.5rem;
.form-wrapper {
   background-color: rgba(190, 190, 190, 0.5);
   width: 100%;
   height: 100%;
   border: 1px dashed black;
.form-wrapper #loading {
```

```
.form-wrapper .upload {
   width: 8rem;
   height: -moz-fit-content;
   height: fit-content;
   border-radius: 6px;
   background-color: rgb(114, 96, 182);
   box-shadow: 0 5px 10px rgb(146, 135, 247);
.form-wrapper .upload #up_btn {
.form-wrapper .upload label {
   font-size: 1rem;
   height: 100%;
   width: 100%;
   padding: 10px;
.form-wrapper .upload svg {
   height: 15px;
   padding-right: 8px;
   margin-bottom: -2px;
@media screen and (max-width: 700px) {
   .upload-container {
       height: 20rem;
       width: 18rem;
       margin-top: 3.5rem;
       margin-bottom: -8rem;
   .heading .heading__main {
       margin-top: -6rem;
       font-size: 2rem;
       padding-bottom: 1rem;
```

### HOME PAGE (JS)

```
feather.replace(); // Load feather icons

form = document.querySelector('.upload')
loading = document.querySelector("#Loading")
select = document.querySelector("#upload-image");

select.addEventListener("change", (e) => {
    e.preventDefault();
    form.submit()
    form.style.visibility = "hidden";
    loading.style.display = 'flex';
});
```

### PREDICT PAGE (HTML)

```
<title>Prediction | Handwritten Digit Recognition</title>
<link rel="stylesheet" href="{{url_for('static',filename='css/predict.css')}}" />
<link rel="icon" type="image/svg" sizes="32x32" href="{{url_for('static',filename='images/icon.svg')}}"</pre>
<meta name="viewport" content="width=device-width, initial-scale=1.0" />
<div class="container">
    <h1>Prediction</h1>
    <div class="result-wrapper">
        <div class="input-image-container">
            <img src="{{url_for('static',filename='data/')}}{{img_name}}" />
        <div class="result-container">
            <div class="value">{{best.0}}</div>
            <div class="accuracy">{{best.1}}%</div>
    <h1>Other Predictions</h1>
    <div class="other_predictions">
        {% for x in others %}
        <div class="value">
            <h2>{{x.0}}</h2>
            <div class="accuracy">{{x.1}}%</div>
```

```
@import url("https://fonts.googleapis.com/css2?family=Overpass:wght@200;300;400;500;600;700;900&display=swap");
   font-family: "Overpass", sans-serif;
   padding-top: 2rem;
.container {
   justify-content: center;
   flex-direction: column;
.result-wrapper {
   width: -webkit-fit-content;
   width: -moz-fit-content;
   height: -moz-fit-content;
   height: fit-content;
   box-shadow: 0 0 10px rgb(126, 125, 125);
   padding: 1.5rem;
   -moz-column-gap: 1rem;
   column-gap: 1rem;
.result-wrapper .input-image-container,
.result-wrapper .result-container {
   width: 15rem;
   height: 15rem;
   border: 1px dashed black;
   justify-content: center;
   flex-direction: column;
   background-color: rgb(209, 206, 206);
```

```
.result-wrapper .input-image-container img {
.result-wrapper .result-container .value {
   font-size: 6rem;
.result-wrapper .result-container .accuracy {
   margin-top: -1rem;
.other_predictions {
   align-items: center;
   flex-wrap: wrap;
   column-gap: 1rem;
   row-gap: 1rem;
.other_predictions .value {
   justify-content: center;
   align-items: center;
   width: 5rem;
   height: 5rem;
   box-shadow: 0 0 7px rgb(158, 157, 157);
.other_predictions .value div {
   margin-top: -1.2rem;
@media screen and (max-width: 700px) {
       font-size: 2.3rem;
   .result-wrapper .input-image-container,
   .result-wrapper .result-container {
       width: 7rem;
       height: 7rem;
   .result-wrapper .result-container .value {
       font-size: 4rem;
```



https://github.com/IBM-EPBL/IBM-Project-1075-1658340258

# **▶** PROJECT DEMO

https://drive.google.com/file/d/1iXRCljJ\_3aVegKCdQHZVOl4BYao\_iSEY/view?usp=share\_link