PROJECT REPORT

ANOVEL METHOD FOR HANDWRITTEN DIGITRECOGNITION SYSTEM

Submitted By

Team ID: PNT2022TMID06491

Team Leader : Kavenesh Team member : Dharani

Team member: Akash

Team member: Nandhakumar

TABLE OF CONTENTS

1. INTRODUCTION

- 1. Project Overview
- 2. Purpose

2. LITERATURE SURVEY

- 1. Existing problem
- 2. References
- 3. Problem Statement Definition

3. IDEATION & PROPOSED SOLUTION

- 1. Empathy Map Canvas
- 2. Ideation & Brainstorming
- 3. Proposed Solution
- 4. Problem Solution fit

4. REQUIREMENT ANALYSIS

- 1. Functional requirement
- 2. Non-Functional requirements

5. PROJECT DESIGN

- 1. Data Flow Diagrams
- 2. Solution & Technical Architecture
- 3. User Stories

6. PROJECT PLANNING & SCHEDULING

- 1. Sprint Planning & Estimation
- 2. Sprint Delivery Schedule
- 3. Reports from JIRA

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

- 1. Feature 1
- 2. Feature 2

3. Database Schema (if Applicable)

8. TESTING

- 1. Test Cases
- 2. User Acceptance Testing

9. RESULTS

1. Performance Metrics

10.ADVANTAGES & DISADVANTAGES

- 11.CONCLUSION
- 12.FUTURE SCOPE
- 13.APPENDIX

Source Code

GitHub & Project Demo Link

1. INTRODUCTION

1.1 Project Overview

Machine learning and deep learning play an important role in computer technology and Artificial Intelligence. With the use of Deep Learning and Machine learning, human effort can be reduced in recognizing, learning, predictions and in many more areas.

Handwritten Digit Recognition is the ability of Computer systems to recognize handwritten digits from various sources, such as images, documents, and so on. This project aims to let users take advantage of machine learning to reduce manual tasks inrecognizing digits.

1.2 Purpose

Digit Recognition system are capable of recognizing the digits from different sources like emails, bank cheques, papers, images, etc. and in different real-world scenarios for online handwriting recognition on computer, tablets or systems, recognizenumber plates of vehicles, processing bank cheque amounts, numeric enteries in formsfilled up by hand (tax forms) and so on.

2. LITERATURE SURVEY

2.1Existing problem

- ✓ The different architectures of CNN, hybrid CNN,CNN RNN and CNNHMM models, and domain specific recognition system, are not thoroughly inquired and evolutionary algorithms are not clearly explored for optimizing CNN learning parameters, the number of layers, learning rate and kernel sizes of convolutional filters.
- ✓ The fluctuation of accuracies for handwritten digits was observed for 15 epochsby varying the hidden layers. There is no clear explanation

given for observing variation in the overall classification accuracy by varying the number of hiddenlayers and batch size.

2.2 References

S.NO	Author Name	Paper Title	Journal/ Conference title	Page No/ Volum eNo	Year of Publicati on	Description
1.	Savita Ahlawat , Amit Choudh ary, Anand Nayyar, Saurabh Singh and Byungu n Yoon.	Improved Handwritten Digit Recognition Using Convolutiona 1 Neural Networks (CNN)	IEEE Sensors Journal		2020	In this paper, with the aim of improving the performance of handwritten digit recognition, they valuated variants of a convolution al neural network to avoid complex preprocessin g, costly feature extraction and a complex ensemble (classifier combination) approach

						of a traditional recognition system.
2.	Vijayala xmi R Rudras wamima th, Bhavani shankar and Channas andra.	Handwritten Digit Recognition using CNN	International Journal of Innovative Science and Research Technology	Volume -4 Issue- 6	2019	In this paper, the most widely used Machine learning algorithms, KNN, SVM, RFC and CNN have been trained and tested on the same data in order acquire the comparison between the classifiers
3.	Fathma Siddiqu e, Shadma n Sakib and Md. Abu Bakr Siddiqu e.	Recognition of Handwritten Digit using Convolutiona l Neural Network in Python with Tensorflow and Comparison of Performance for Various Hidden Layers	5th International Conference on Advances in Electrical Engineering (ICAEE)		2019	In this paper, they observed the variation of accuracies of CNN to classify handwritten digits for 15 epochs using various numbers of hidden layers and epochs and

			to make the comparison between the
			accuracies.
			For this
			performance
			evaluation
			of CNN,
			they
			performed
			the
			experiment
			using
			Modified
			National
			Institute of
			Standards
			and
			Technology(
			MN IST)
			dataset.

4.	Akanks	Review on	International	Volume	2021	In this
••	ha		Journal of	-9 Issue-	2021	
		Deep				paper,
	Gupta,	Learning	Recent	5		Object
	Ravindr	Handwritten	Technology			Character
	a Pratap	Digit	and			Recognition
	Narwari	Recognition	Engineering			(OCR) is
	a and	using	(IJRTE)			used on
	Madhav	Convolutiona				printed or
	Singh	1 Neural				documented
		Network				letters to
						convert
						them into
						text. The
						database has
						training
						image
						database of
						60,000
						images and

			testing
			image
			database of
			10,000
			images. The
			KNN
			algorithm
			describes
			categorical
			value by
			making use
			of majority
			of votes of
			K - nearest
			neighbors,
			the K value
			used to
			differ here.

5.	Md.	Recognition	Global	Volume	2019	The goal of
	Anwar	of	Journal of	19	_017	this work
	Hossain	Handwritten	Computer	Issue2		will be to
	and Md.	Digit using	Science and			create a
	Mohon	Convolutiona	Technology:			model that
	Ali	l Neural	D Neural &			will be able
		Network	Artificial			to identify
		(CNN)	Intelligence			and
						determine
						the
						handwritten
						digit from
						its image
						with better
						accuracy
						using using
						the concepts
						of
						Convolution
						al Neural
						Network
						and MNIST

			dataset.
			Later it can
			be extended
			for character
			recognition
			and real-
			time
			person's
			handwriting.
			The results
			can be made
			more
			accurate
			with more
			convolution
			layers and
			more
			number of
			hidden
			neurons.

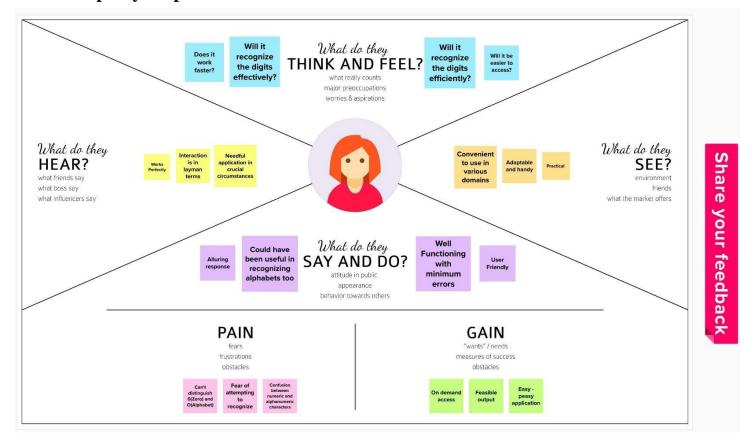
2.3 Problem Statement Definition

Handwriting recognition is one of the compelling research works going on because every individual in this world has their own style of writing. It is the capability of the computer to identify and understand handwritten digits or characters automatically. Because of the progress in the field of science and technology, everything is being digitalized to reduce human effort.

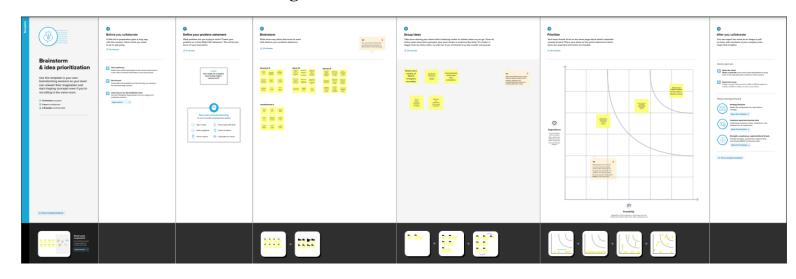
Hence, there comes a need for handwritten digit recognition in many real time applications. MNIST data set is widely used for this recognition process and it has 70000 handwritten digits. We use Artificial neural networks to train these images and build a deep learning model. Web application is created where the user can upload an image of a handwritten digit. this image is analyzed by the model and the detected result is returned on to UI(User Interface).

3. IDEATION & PROPOSED SOLUTION

3.1 Empathy Map Canvas



3.2 Ideation & Brainstorming



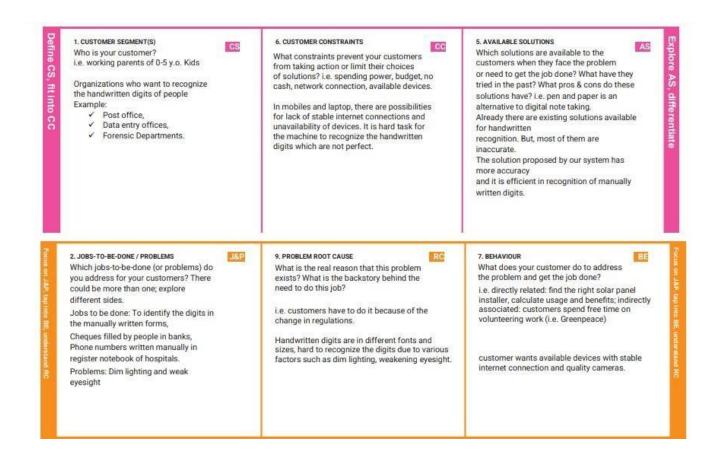
3.3 Proposed Solution

S.No.	Parameter	Description
1	Problem Statement (Problem to be solved)	Handwriting recognition is one of the compelling research works going on because every individual in this world has their own style of writing. It is the capability of the computer to identify and understand handwritten digits or characters automatically. Because of the progress in the field of science and technology, everything is being digitized to reduce humaneffort. Hence, there comes a need for handwritten digit recognition in many real-time applications. The user interacts with the UI (User Interface) to upload the image as input. The uploaded image is analyzed by the model which is integrated. Once the model analyses the uploaded image, the predictionis showcased on the UI.
2	Idea / Solution description	Convolutional Neural Networks(CNN) has become one of the most appealing approaches andhas been an ultimate factor in a variety of recent success and challenging machine learning applications. In our model we use AlexNet, which is one of the CNN architectures. AlexNet allows for multi-GPUtraining by putting half of the model's neurons on one GPU and the other half on another

		GPU. Not only does this mean that a bigger model can be trained, but it also cuts down onthe training time. It also reduces the overfitting problem by DataAugmentation and Dropout.
3	Novelty / Uniqueness	Handwritten Digit Recognition is the capability of a computer to fete the mortal handwritten integers from different sources like images, papers, touch defenses, etc. And classify theminto 10 predefined classes (0-9). This is the existing method along with this we add some features to make our project unique among them.
4	Social Impact / Customer Satisfaction	Even the unclear or blurred digits can be recognized after the removal of noise and data preprocessing. One such application is a handwritten digit recognition system that can be used in postal mail sorting, bank check processing, form data entry, etc.,
5	Business Model (Revenue Model)	Handwritten digit recognition isnecessary because everything isdigitalized. The benefits of handwritten digit recognizer is high. In the banking sector, it isvery efficient. It is used to recognize the figures written oncheques. So, Varied handwriting of each and every person in the cheque can be identified. Handwritten addresses are difficult to sort by machine, not

		necessarily because of sloppy handwriting, but because peoplewrite all over the envelope. We have hard time segmenting handwritten addresses into their components, such as ZIP code or street address, because very few people print addresses neatly in a prescribed format. So, this problem can be solvedusing Handwritten digit recognition system.
6	Scalability of the Solution	In our model, AlexNet significantly outperformed as it is trained on a GTX 580 GPU with only 3 GB of memory which couldn't fit the entire network. So the network was split across 2 GPUs, with half of the neurons(feature maps) oneach GPU. So, a greater accuracy can be attained by allowing multi-GPU training byputting half of the model's neurons on one GPU and the other half on another GPU.

3.4 Problem Solution fit



3. TRIGGERS What triggers customers to act? i.e. seeing their neighbour installing solar panels, reading about a more efficient solution in the news. Advertisement in the market about the efficient recognition of digits. Articles about the achievements made by our project.	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. Our solution aims to recognize handwritten digits using machine learning techniques thereby saving costs to the organization improving	8. CHANNELS of BEHAVIOUR 8.1 ONLINE What kind of actions do customers take online? Extract online channels from #7 Requires Stable internet connection for image processing. 8.2 OFFLINE What kind of actions do customers take offline? Extract offline channels from #7 and use them for customer development.
4. EMOTIONS: BEFORE / AFTER How do customers feel when they face a problem or a job and afterwards? i.e. lost, insecure > confident, in control - use it in your communication strategy & design. Defects are common and our project is not an exception When the system failed to recognize the digit, Customer Mentality: Before:(Failure) We would give guarantee that it would work most of the time	employee productivity. In our model we use AlexNet , which is one of the CNN architectures . AlexNet allows for multi-GPU training by putting half of the model's neurons on one GPU and the other half on another GPU. Not only does this mean that a bigger model can be trained, but it also cuts down on the training time. It also reduces the overfitting problem by Data Augmentation and Dropout.	Obtain modern electronic devices and check they are working
and if any error occurs, they can contact us at any time. So, customers can feel at ease. After:(Failure) They have no need to panic when the failure occurs They can easily contact us to rectify the error. We would solve the defect as soon as possible.		

4. REQUIREMENT ANALYSIS

4.3 Functional requirement

Following are the functional requirements of the proposed solution.

FR No.	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)
FR-1	Input Correlation	Digital image correlation is a technique that combines image registration and tracking methods for accurate 2D measurements of changes in images and recognizes the characters from the images.
FR-2	Data Preparation	Data preparation is the process of preparing raw data so that it is suitable for further processing and analysis.
FR-3	Feature Extraction	Feature extraction refers to the process of transforming raw data into numerical features that can be processed while preserving the information in the original data set.
FR-4	Character Classification	In character classification phase, the attributes of the data in the picture are compared to the classes in the database to determine in which class the picture belongs to.

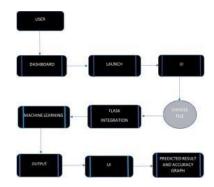
4.4 Non-Functional requirements

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

FR No.	Non-Functional Requirement	Description				
NFR-1	Usability	Handwritten digit recognition is one of the major important issues in pattern recognition applications. Some of the applications for digit recognition include data entry forms, Bank check processing etc,.				
NFR-2	Security	The applications of handwritten digit recognition can be used in the banking sector where it can be used to maintain the security pin numbers safely. It can be also used for blind-people by using sound output.				
NFR-3	Reliability	Reliability indicates the probability that the system will perform its intended function for a larger period of sufficient time and also it will operate in a secured environment without any failures.				
NFR-4	Performance	The standard implementations of neural networks achieve an accuracy of approximately (98–99)				
		percent in correctly classifying the handwritten digits.				
NFR-5	Availability	The features for handwritten digit recognition have been Acquainted. These features are based on shape analysis of the digit image and extract slant or slope information. They are effective in obtaining good recognition of accuracy.				
NFR-6	Scalability	The scalability in the task of handwritten digit recognition, using a classifier, has great importance and it makes use of online handwriting recognition on computer tablets, recognizing zip codes on mail for postal mail sorting, processing bank check amounts, numeric entries in forms filled up manually(for example - tax forms) and so on.				

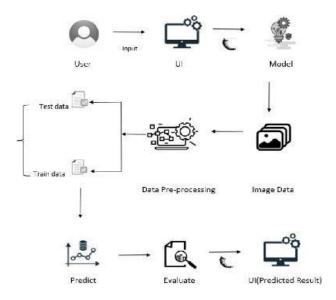
5. PROJECT DESIGN

5.1Data Flow Diagrams



5.2 Solution & Technical Architecture

Solution Architecture



Technology Architecture

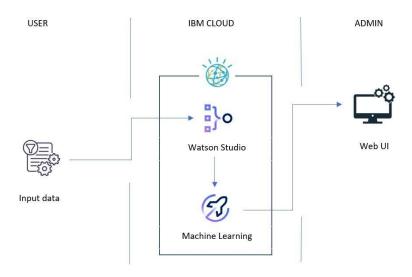


Table-1 : Components & Technologies:

S.No	Component	Description	Technology HTML, CSS, JavaScript		
1.	User Interface	How user interacts with application e.g. Web UI			
2.	Application Logic-1	Model is built	Python		
3.	Application Logic-2	Python model is deployed	IBM Watson Studio		
4.	File Storage	Predicted outputs of the image are stored in a local folder.	Local Filesystem		
5.	Machine Learning Model	To predict the image uploaded by the user.	Image Recognition Model		
6.	Infrastructure (Server / Cloud)	Application Deployment on Local System / Cloud Local Server Configuration: Flask Cloud Server Configuration: IBM Watson Studio	Local, Cloud Foundry.		

Table-2: Application Characteristics:

S.No	Characteristics	Description	Technology
1.	Open-Source Frameworks	List the open-source frameworks used	Flask
2.	Security Implementations	List all the security / access controls implemented, use of firewalls etc.	e.g. SHA-256, Encryptions, IAM Controls, OWASP etc.
3.	Scalable Architecture	High workload can be supported without undergoing any major changes.	Technology used in the architecture is that with Python and the IBM cloud.
4.	Availability	Readily available enables the IT Infrastructure to function when some of the components fail.	Technology used is IBM cloud.
5.	Performance	Performance technology is a field which uses various tools,processes and procedures in a systematic and efficient manner to improve the desired outcomes of individuals and organizations.	Technology used is python.

5.3User Stories

User Stories

Use the below template to list all the user stories for the product.

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer	Dashboard	USN-1	As a user, they can see the information regarding the prediction of handwritten digit recognition.	I can see the information regarding digit recognition.	High	Sprint 1
2)	Launch	USN-2	On clicking the launch button, it will redirect the user to a page where the images to be predicted can be uploaded.	I can see the launch button.	High	Sprint 1
	Upload	USN-3	Users can select the image from the local storage.	I can upload the image.	High	Sprint 2
	Predict	USN-4	Once the image is uploaded, it will predict the respective image.		High	Sprint 3
	Display	USN-5	The predicted image will be displayed with the accuracy chart.	I can see the result with accuracy.	High	Sprint 4

6. PROJECT PLANNING & SCHEDULING

6.1Sprint Planning & Estimation

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Dashboard	USN-1	As a user, they can see the information	2	High	Pavithrah M,
			regarding the prediction of handwritten digit			Nandhini S,
			recognition.			Lakshmi A,
						Visaka L
Sprint-1	Launch	USN-2	On clicking the launch button, it will redirect the	2	High	Pavithrah M,
			user to a page where the images to be			Nandhini S,
			predicted can be uploaded.			Lakshmi A,
						Visaka L
Sprint-2	Upload	USN-3	Users can select the image from the local	2	High	Nandhini S,
	2000		storage.		220	Visaka L
Sprint-3	Predict	USN-4	Once the image is uploaded, it will predict the	2	High	Lakshmi A,
			respective image.		5004	Pavithrah M
Sprint-4	Display	USN-5	The predicted image will be displayed with the	2	High	Pavithrah M,
	370 34		accuracy chart.		0.000.00	Nandhini S,
			*			Lakshmi A,
						Visaka L

6.2 Sprint Delivery Schedule

Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	20	6 Days	24 Oct 2022	29 Oct 2022	20	29 Oct 2022
Sprint-2	20	6 Days	31 Oct 2022	05 Nov 2022	20	05 Nov 2022
Sprint-3	20	6 Days	07 Nov 2022	12 Nov 2022	20	12 Nov 2022
Sprint-4	20	6 Days	14 Nov 2022	19 Nov 2022	20	19 Nov 2022

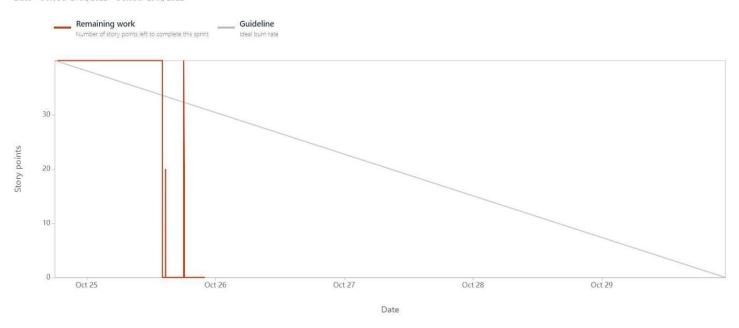
6.3 Reports from JIRA

Velocity Report



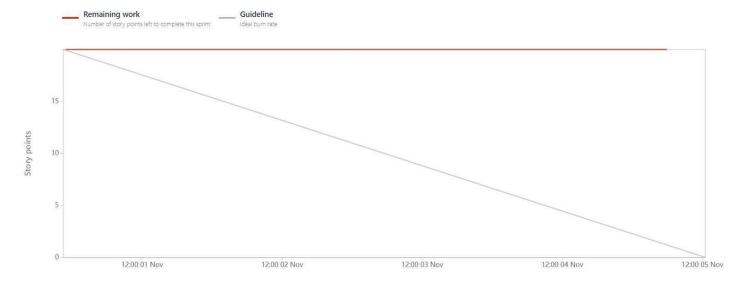
Sprint 1

Date - October 24th, 2022 - October 29th, 2022



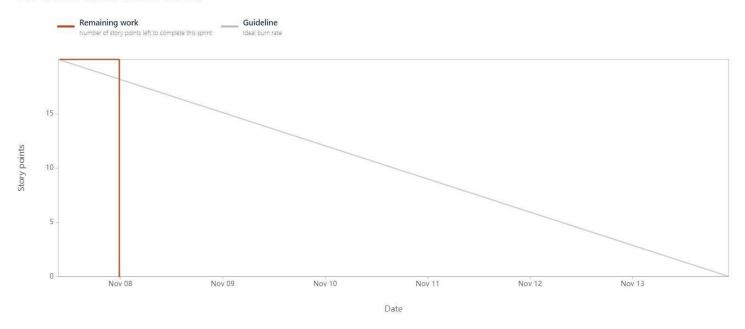
Sprint 2

Date - October 31st, 2022 - November 5th, 2022



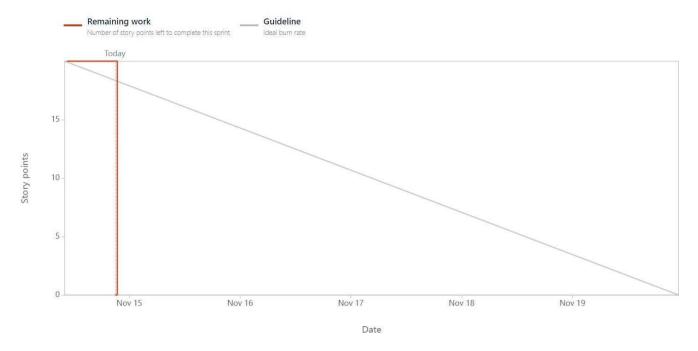
Sprint 3

Date - November 7th, 2022 - November 13th, 2022



Sprint 4

Date - November 14th, 2022 - November 19th, 2022



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

```
app.py
```

```
import numpy as np
import os
from PIL import Image
from flask import Flask, request, render_template, url_for
from werkzeug.utils import secure_filename, redirect
from gevent.pywsgi import WSGIServer
from keras.models import load_model
from keras.preprocessing import image
from flask import send_from_directory
UPLOAD FOLDER = r'C:\Users\ELCOT\Documents\GitHub\IBM-Project-27149-
1660047584\Final Deliverables\Final code\app'
app = Flask( name )
app.config['UPLOAD_FOLDER'] = UPLOAD_FOLDER
model = load_model("mnistCNN.h5")
@app.route('/')
def index():
  return render template('index.html')
@app.route('/predict', methods=['GET', 'POST'])
def upload():
  if request.method == "POST":
    f = request.files["image"]
    filepath = secure_filename(f.filename)
    f.save(os.path.join(app.config['UPLOAD_FOLDER'], filepath))
    upload_img = os.path.join(UPLOAD_FOLDER, filepath)
                                         25
```

```
img = Image.open(upload_img).convert("L") # convert image to monochrome
img = img.resize((28, 28)) # resizing of input image

im2arr = np.array(img) # converting to image
im2arr = im2arr.reshape(1, 28, 28, 1) # reshaping according to our requirement

pred = model.predict(im2arr)

num = np.argmax(pred, axis=1) # printing our Labels

return render_template('predict.html', num=str(num[0]))

if __name__ == '__main__':
    app.run(debug=True, threaded=False)
```

8. TESTING

8.1 Test Cases

Test case ID	Feature Type	Component	Test Scenario	Expected Result	Actual Result	Status
Homepage_TC_OO1	Functional	Home Page	Verify user is able to see the Homepage when clicked on the link	Home Page should be displayed.	Working as expected	Pass
Homepage_TC_OO2	UI	Home Page	Verify the UI elements in Homepage	Application should show below UI elements: a.choose file button b.predict button c.clear button	Working as expected	Pass
Homepage_TC_OO3	Functional	Home Page	Verify user is able to choose file from the local system and click on predict	Choose file popup screen must be displayed and user should be able to click on predict button	Working as expected	Pass
Homepage_TC_OO4	Functional	Home page	Verify user able to select invalid file format	Application won't allow to attach formats other than ".png, .jiff, .pjp, .jpeg, .jpg, .pjpeg"	Working as expected	Pass
Predict_TC_OO5	Functional	Predict page	Verify user is able to navigate to the predict to and view the predicted result	User must be navigated to the predict page and must view the predicted result	Working as expected	Pass

8.2 User Acceptance Testing

Defect Analysis

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Subtotal
By Design	0	0	0	0	0
Duplicate	0	0	0	0	0
External	0	0	0	0	0
Fixed	0	0	0	0	0
Not Reproduced	0	0	0	0	0
Skipped	0	0	0	0	0
Won't Fix	0	0	0	0	0
Totals	0	0	0	0	0

Test Case Analysis

Section	Total Cases	Not Tested	Fail	Pass
Client Application	5	0	0	5
Security	5	0	0	5
Final Report Output	5	0	0	5
Performance	5	0	0	5

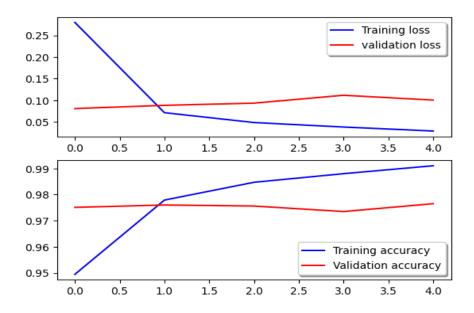
9. RESULTS

9.1 Performance Metrics

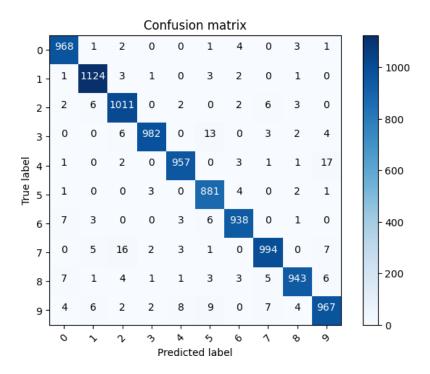
Model Summary:

Model: "sequential"		
Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 26, 26, 64)	640
conv2d_1 (Conv2D)	(None, 24, 24, 32)	18464
flatten (Flatten)	(None, 18432)	0
dense (Dense)	(None, 10)	184330
Total params: 203,434		-=====
Trainable params: 203,434		
Non-trainable params: 0		
None		

Accuracy:



Confusion Matrix:

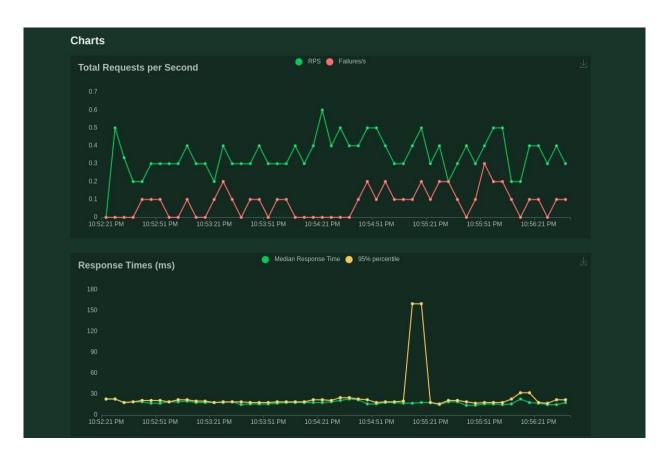


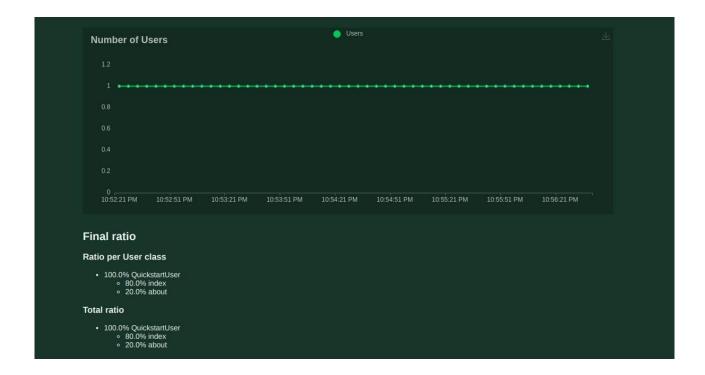
Classification Report:

	precision	recall	f1-score	support	
0	0.98	0.99	0.98	980	
1	0.98	0.99	0.99	1135	
2	0.97	0.98	0.97	1032	
3	0.99	0.97	0.98	1010	
4	0.98	0.97	0.98	982	
5	0.96	0.99	0.97	892	
6	0.98	0.98	0.98	958	
7	0.98	0.97	0.97	1028	
8	0.98	0.97	0.98	974	
9	0.96	0.96	0.96	1009	
accuracy			0.98	10000	
macro avg	0.98	0.98	0.98	10000	
weighted avg	0.98	0.98	0.98	10000	

Performance Metrics Result:

Locust Test Report										
During: 11/15/2022, 10:52:19 PM - 11/15/2022, 10:56:36 PM										
Target Host: http://127.0.0.1:5000/										
Script: locustfile.py										
Request Statistics										
Method	Name	# Requests	# Fails	Average (ms)	Min (ms)	Max (ms)	Average size (b	ytes)	RPS	Failures/s
GET		67		17	12	24	5875		0.3	0.0
GET	//predict	23	23	21	11	163	265		0.1	0.1
	Aggregated	90	23	18	11	163	4441		0.4	0.1
Respor	ise Time Si	tatistics								
Method	Name	50%ile (ms)	60%ile (ms)	70%ile (ms)	80%ile (ms)	90%ile (ms)	95%ile (ms)	99%ile (n	ns)	100%ile (ms)
GET		18	18	19	19	22	23	25		25
GET	//predict	15	15	16	16	17	32	160		160
	Aggregated	17	18	18	19	22	23	160		160





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 high performance server for faster predictions.
- ✓ Prone to occasional errors.

11. CONCLUSION

This project demonstrated a web application that uses machine learning to recognie 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 number plates 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.

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.

13.APPENDIX

Source Code HTML AND CSS:

index.html:

```
<html>
<head>
 <title>Handwritten Digit Recognition</title>
 <meta name="viewport" content="width=device-width">
 <!-- GoogleFont -->
 link
href="https://fonts.googleapis.com/css2?family=Prompt:wght@600&display=swap"
rel="stylesheet">
 <link href="https://fonts.googleapis.com/css2?family=Varela+Round&display=swap"</pre>
rel="stylesheet">
 link
href="https://fonts.googleapis.com/css2?family=Source+Code+Pro:wght@500&display
=swap" rel="stylesheet">
 link
href="https://fonts.googleapis.com/css?family=Calistoga|Josefin+Sans:400,700|Pacifico
&display=swap" rel="stylesheet">
 <!-- bootstrap -->
 k rel="stylesheet"
href="https://stackpath.bootstrapcdn.com/bootstrap/4.3.1/css/bootstrap.min.css"
integrity="sha384-
ggOyR0iXCbMQv3Xipma34MD+dH/1fQ784/j6cY/iJTQUOhcWr7x9JvoRxT2MZw1T
" crossorigin="anonymous">
 <link rel="stylesheet" type= "text/css" href= "{{</pre>
url_for('static',filename='css/style.css') }}">
 <!-- fontawesome -->
 <script src="https://kit.fontawesome.com/b3aed9cb07.js"</pre>
crossorigin="anonymous"></script>
```

```
<script src="https://code.jquery.com/jquery-3.3.1.slim.min.js" integrity="sha384-</pre>
q8i/X+965DzO0rT7abK41JStQIAqVgRVzpbzo5smXKp4YfRvH+8abtTE1Pi6jizo"
crossorigin="anonymous"></script>
 <script
src="https://cdnjs.cloudflare.com/ajax/libs/popper.js/1.14.7/umd/popper.min.js"
integrity="sha384-
UO2eT0CpHqdSJQ6hJty5KVphtPhzWj9WO1clHTMGa3JDZwrnQq4sF86dIHNDz0W
1" crossorigin="anonymous"></script>
 <script src="https://stackpath.bootstrapcdn.com/bootstrap/4.3.1/js/bootstrap.min.js"</pre>
integrity="sha384-
JjSmVgyd0p3pXB1rRibZUAYoIIy6OrQ6VrjIEaFf/nJGzIxFDsf4x0xIM+B07jRM"
crossorigin="anonymous"></script>
 <script src="https://cdn.jsdelivr.net/npm/@tensorflow/tfjs@latest"></script>
</head>
<script>
 function preview() {
  frame.src=URL.createObjectURL(event.target.files[0]);
}
  $(document).ready(function() {
     $('#clear_button').on('click', function() {
        $('#image').val(");
        $('#frame').attr('src',"");
       });
    });
```

```
</script>
<body>
 <h1 class="welcome">Handwritten Digit Recognition Application <br>
 <div id="team_id">TEAM ID : PNT2022TMID06491</div>
 </h1>
 <section id="title">
  <br>><br>>
   >
    A Novel Method for Handwritten Digit Recognition System.
   This is a very useful application as it allows you to quickly jot down numbers
for contacts, postal mail sorting, bank check processing, form data entry, etc.
 </section>
 <section id="content">
    <div class="leftside">
    <form action="/predict" method="POST" enctype="multipart/form-data">
    <label style="color: #000;">Select a image:</label>
    <input id="image" type="file" name="image" accept="image/png, image/jpeg"</pre>
onchange="preview()"><br><br>
     <img id="frame" src="" width="100px" height="100px"/>
     <div class="buttons_div">
```

```
<button type="submit" class="btn btn-dark"
 id="predict_button">Predict</button>
        <button type="button" class="btn btn-dark" id="clear_button">&nbsp Clear
 &nbsp</button>
       </div>
      </form>
      </div>
      <br>>
  </section>
 </body>
 </html>
Style.css
#clear_button{
 margin-left: 15px;
 font-weight: bold;
 color: #153462;
}
#confidence{
 font-family: 'Josefin Sans', sans-serif;
 margin-top: 7.5%;
}
#content{
 margin: 0 auto;
 padding: 2% 15%;
 padding-bottom: 0;
 background-color: #4FA095;
```

```
.welcome{
 text-align: center;
 position: relative;
 color: #153462;
 background-color: #F6F6C9;
 padding-top: 1%;
 font-weight: bold;
 font-family: 'Prompt', sans-serif;
#team_id{
 text-align: right;
 font-size: 25px;
 padding-right: 3%;
}
#predict_button{
 margin-right: 15px;
 color: #153462;
 font-weight: bold;
#prediction_heading{
 font-family: 'Josefin Sans', sans-serif;
 margin-top: 7.5%;
```

}

```
#result{
 font-size: 5rem;
}
#title{
 padding: 1.5% 15%;
 margin: 0 auto;
 text-align: center;
 background-color: #BAD1C2;
.btn {
  font-size: 15px;
  padding: 10px;
  -webkit-appearance: none;
  background: #eee;
  border: 1px solid #888;
  margin-top: 20px;
  margin-bottom: 20px;
  border-radius: 25px;
  color: #4FA095;
}
.buttons\_div\{
 margin-bottom: 30px;
 margin-right: 80px;
```

```
.heading{
 font-family: 'Varela Round', sans-serif;
 font-weight: 700;
 font-size: 2rem;
 display: inline;
}
.leftside{
 text-align: center;
 margin: 0 auto;
 margin-top: 2%;
 /* padding-left: 10%; */
}
#frame{
 margin-right: 10%;
}
.predicted_answer{
 text-align: center;
 margin: 0 auto;
 padding: 3% 5%;
 padding-top: 0;
 /* padding-left: 10%; */
}
p{
```

```
font-family: 'Source Code Pro', monospace, sans-serif;
 margin-top: 1%;
 color: #153462;
}
@media (min-width: 720px) {
 .leftside{
  padding-left: 10%;
 }
}
Predict.html
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <title>Prediction</title>
</head>
<style>
  body{
  background-color: #BAD1C2;
  }
  #rectangle{
  width:500px;
  height:120px;
  background-color: #4FA095;
```

```
border-radius: 25px;
  position:absolute;
  top:25%;
  left:50%;
  transform:translate(-50%,50%);
  }
  #ans{
 text-align: center;
 font-size: 40px;
 margin: 0 auto;
 padding: 3% 5%;
 padding-top: 8%;
 color: #000;
  }
</style>
<body>
  <div id="rectangle">
    <h1 id="ans">Predicted Number is {{num}}</h1>
  </div>
</body>
</html>
```

MODEL CREATION:

IMPORTING THE REQUIRED LIBRARIES

import numpy as np

import tensorflow #open source used for both ML and DL for computation

from tensorflow.keras.datasets import mnist #mnist dataset

from tensorflow.keras.models import Sequential #it is a plain stack of layers

from tensorflow.keras import layers #A Layer consists of a tensor- in tensor-out computat ion funct ion

from tensorflow.keras.layers import Dense, Flatten #Dense-Dense Layer is the regular deeply connected r

#faltten -used fot flattening the input or change the dimension

from tensorflow.keras.layers import Conv2D #onvoLutiona l Layer

from keras.optimizers import Adam #opt imizer

from keras. utils import np_utils #used for one-hot encoding

import matplotlib.pyplot as plt #used for data visualization

(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 #storing the no of classes in a variable

y_train = np_utils.to_categorical (y_train, number_of_classes) #converts the output in binary format

y_test = np_utils.to_categorical (y_test, number_of_classes)

ADD CNN LAYER

#create model model=Sequential ()

```
#adding modeL Layer
model.add(Conv2D(64, (3, 3), input_shape=(28, 28, 1), activation='relu'))
model.add(Conv2D(32, (3, 3), activation = 'relu'))
#flatten the dimension of the image
model.add(Flatten())
#output layer with 10 neurons
model.add(Dense(number_of_classes,activation = 'softmax'))
```

COMPILE THE MODEL

```
#Compile model
model.compile(loss='categorical crossentropy', optimizer="Adam", metrics=['accuracy'])
x_{train} = np.asarray(x_{train})
y_train = np.asarray(y_train)
TRAIN THE MODEL
#fit the model
model.fit(x train, y train, validation data=(x test, y test), epochs=5, batch size=32)
Epoch 1/5
accuracy: 0.9508 - val_loss: 0.1043 - val_accuracy: 0.9667
Epoch 2/5
accuracy: 0.9778 - val_loss: 0.0785 - val_accuracy: 0.9769
Epoch 3/5
accuracy: 0.9844 - val_loss: 0.1036 - val_accuracy: 0.9697
Epoch 4/5
accuracy: 0.9895 - val_loss: 0.0911 - val_accuracy: 0.9763
Epoch 5/5
accuracy: 0.9902 - val loss: 0.0907 - val accuracy: 0.9790
```

OBSERVING THE METRICS

SAVE THE MODEL

Save the model model.save('models/mnistCNN.h5')

GitHub

GitHub Link

https://github.com/IBM-EPBL/IBM-Project-27149-1660047584

Demo Video

https://drive.google.com/file/d/1Cz5G2PN_HJx6EremoLFpncffZfS_3k5S/view?usp=sharing

https://github.com/IBM-EPBL/IBM-Project-27149-1660047584/tree/main/Final%20Deliverables