

PROJECT REPORT

A NOVEL METHOD FOR HANDWRITTEN DIGIT RECOGNITION

submitted by

PNT2022TMID05025

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CHAPTER 1

INTRODUCTION

1.1 PROJECT OVERVIEW

Handwritten Digit Recognition is the ability of computer systems to recognise 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 in recognizing digits.

The 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 the UI. In the Banking Sector too where more handwritten numbers are involved like account number, figure of cash and checks.

1.2 PURPOSE

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

CHAPTER 2

LITERATURE SURVEY

2.1 EXISTING PROBLEM

The fundamental problem with handwritten digit recognition is that handwritten digits do not always have the same size, width, orientation, and margins since they vary from person to person. Additionally, there would be issues with identifying the numbers because of similarities between numerals like 1 and 7, 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

1. FaisalTehseen Shah, Kamran Yousaf,"Handwritten Digit Recognition Using Image Processing and Neural Networks", Proceedings of the World Congress on Engineering, vol., 2007.
2. Dr.Kusumgupta2 ,"a comprehensive review on handwritten digit recognition using various neural network approaches", international journal of enhanced research in management & computer applications, vol. 5, no. 5, pp. 22-25, 2016.
3. Saeed AL-Mansoori,"Intelligent Handwritten Digit Recognition using Artificial Neural Network", Int. Journal of Engineering Research and Applications, vol. 5, no. 5, pp. 46-51, 2015.
4. Viragkumar N. Jagtap , Shailendra K. Mishra,"Fast Efficient Artificial Neural Network for Handwritten Digit Recognition", International Journal of Computer Science and Information Technologies, vol. 52, no. 0975- 9646, pp. 2302-2306, 2014.

5. Ishani Patel, ViragJagtap and OmpriyaKale."A Survey on Feature Extraction Methods for Handwritten Digits Recognition", International Journal of Computer Applications, vol. 107, no. 12, pp. 11-17, 2014.
6. Dutt A, Dutt A (2017) Handwritten digit recognition using deep learning. Int J AdvRes Comput Eng Technol 6(7):990–997

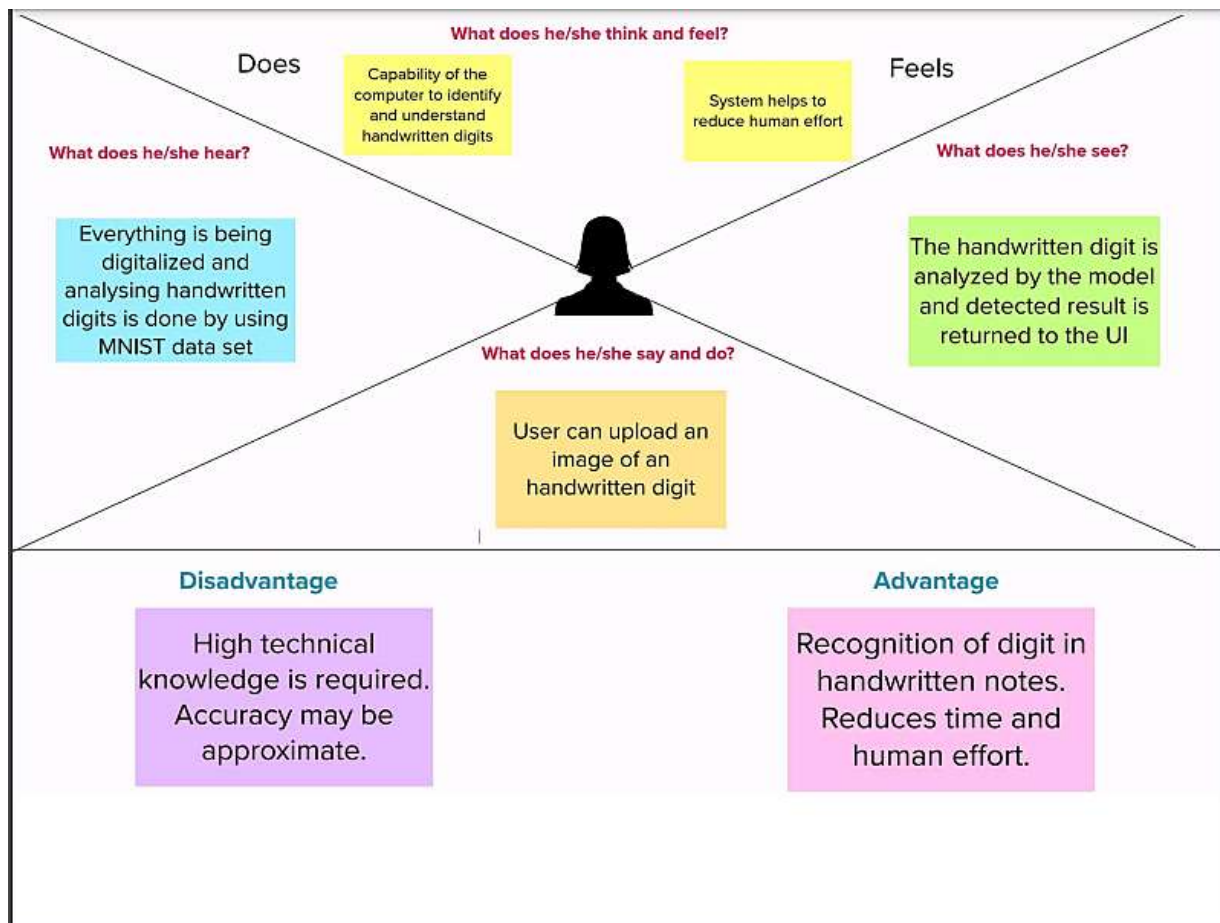
2.3 PROBLEM STATEMENT DEFINITION

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. The 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 the UI. In the Banking Sector too where more handwritten numbers are involved like account number, figure of cash and checks.

CHAPTER 3


IDEATION AND PROPOSED SOLUTION

3.1 EMPATHY MAP CANVAS



3.2 IDEATION & BRAINSTORMING

Template



Brainstorm & idea prioritization

Use this template in your own brainstorming sessions so your team can unleash their imagination and start shaping concepts even if you're not sitting in the same room.

10 minutes to prepare

1 hour to collaborate

2-8 people recommended

Share template feedback

4

Before you collaborate

A little bit of preparation goes a long way with this session. Here's what you need to do to get going.

10 minutes

Team gathering

Set the goal

Learn how to use the facilitation tools

Open article

1

Define your problem statement

What problem are you trying to solve? Frame your problem as a How Might We statement. This will be the focus of your brainstorm.

5 minutes

PROBLEM

To identify the letters of the sentence written by the user in their devices and to convert the handwritten text into digital format.

2

Key rules of brainstorming

To run an smooth and productive session

Stay in topic

Defer judgement


Go for volume

Encourage wild ideas

Listen to others

If possible, be visual

Recognition - driven segmentation is a capability of the generative models. Training for the procedure is relatively simple and quick because there are just a few parameters involved



Need some inspiration?

See a detailed version of this template in Action! per week.

Open example

3.3 PROPOSED SOLUTION

S.No.	Parameter	Description
1.	Problem Statement (Problem to be solved)	To recognize the handwritten digits from a wide variety of sources like emails, papers, images, letters etc. It can solve more complex problems and makes humans' jobs easier.
2.	Idea / Solution description	This system is built by using (CNN) Convolutional Neural Network. By using this system, we can capture the image of handwritten digits and can predict the digits.
3.	Novelty / Uniqueness	Here users can upload the handwritten digits from anywhere and it gives accurate results.
4.	Social Impact / Customer Satisfaction	The feasibility of implementing this idea is moderate neither easy nor tough because the system needs to satisfy the basic requirements of the customer and should give accurate results.
5.	Business Model (Revenue Model)	By using this website, the users can predict and analyze the handwritten digits of the user. The website can be developed at minimum cost with high performance and interactive user interface.
6.	Scalability of the Solution	The solution can be made scalable. The people can easily understand the digits. This system can also be integrated with future technologies

3.4 PROBLEM SOLUTION FIT

Define CS, fit into CC	1. CUSTOMER SEGMENT(S) CS <ul style="list-style-type: none"> Clients are those who use handwritten numbers at work or in institutions like banks, colleges, trains, etc. 	6. CUSTOMER CONSTRAINTS CC <ul style="list-style-type: none"> Internet connections that are unreliable, lack of laptops and mobile, and the inaccessibility of cameras that are suitable. Computers have a difficult time processing handwritten numbers since they are frequently inaccurate and can have a wide range of preferences. The use of handwritten digit recognition, problem can be resolved by identifying the digit that is present in a picture 	5. AVAILABLE SOLUTIONS AS <ul style="list-style-type: none"> Despite the fact that there are currently alternatives to this strategy, they are not very accurate, reliable, or rotation- and variation-invariant. A computer's capacity to respect the imperfect handwriting found in a variety of contexts, such as in documents, pictures, and touch input. 	Explore AS, differentiate
	2. JOBS-TO-BE-DONE / PROBLEMS J&P <ul style="list-style-type: none"> Understanding and analyzing the scribbled numbers is really difficult. More training data are needed. Dim illumination, poor eyesight, difficult to distinguish the digits. 	9. PROBLEM ROOT CAUSE RC <ul style="list-style-type: none"> Due to several issues including deteriorating eyesight, a lack of time, etc., handwritten digits are more challenging to read because they are written in a variety of fonts and size. (Example) customers have to do it because of the change in regulations. 	7. BEHAVIOUR BE <ul style="list-style-type: none"> Finding the software that recognizes numbers most fast and accurately. The client requests dependable internet and top-notch cameras. 	
Identify strong TR & EM	3. TRIGGERS TR <ul style="list-style-type: none"> One of the easiest methods to communicate with a computer and learn the language is through information exchange, which is made simple. Accurately and efficiently gathering the data. 	10. YOUR SOLUTION SL <ul style="list-style-type: none"> The Convolutional Neural Network (CNN) method is used in the solution to accurately recognize handwritten digits. enhancing worker productivity and decreasing costs for the business. 	8. CHANNELS of BEHAVIOUR CH <p>OFFLINE Invest in modern electronics and make sure they work.</p> <p>ONLINE A reliable internet connection is needed for both photo processing and uploading.</p>	Identify strong TR & EM
	4. EMOTIONS: BEFORE / AFTER EM <p>Before: Unsure, Protected, and Confused. AFTER: Confident, upright, and reasonable</p>			

CHAPTER 4

REQUIREMENT ANALYSIS

4.1 FUNCTIONAL REQUIREMENTS

FR No.	Functional Requirement (Epic)	Sub Requirement (Story/ Sub-Task)
FR-1	MNIST Dataset	A training set of 60,000 instances and a test set are included in the modified National Institute of Standards and Technology dataset (MNIST) database of handwritten digits. 10,000 examples in a set.
FR-2	Data pre-processing	Enhances the image by applying a few adjustments to the input image to get it ready for segmentation.
FR-3	Image Data	Handwritten digit recognition refers to a computer's capacity to identify human handwritten digits from a variety of sources, such as photographs, documents, touch screens, etc., and categorise them into ten established classifications (0-9). In the realm of deep learning, this has been the subject of countless studies
FR-4	Website	Web hosting makes the code, graphics, and other items that make up a website accessible online. A server hosts every website you have ever visited. The type of hosting determines how much space is allotted to a website on a server. Shared, dedicated, VPS, and reseller hosting are the four basic varieties.
FR-5	Digit Classifier Model	To train a convolutional network to predict the digit from an image, use the MNIST database of handwritten digits. get the training and validation data first
FR-6	Evaluation	Make that the model recognizes the digit correctly and generates the correct result.

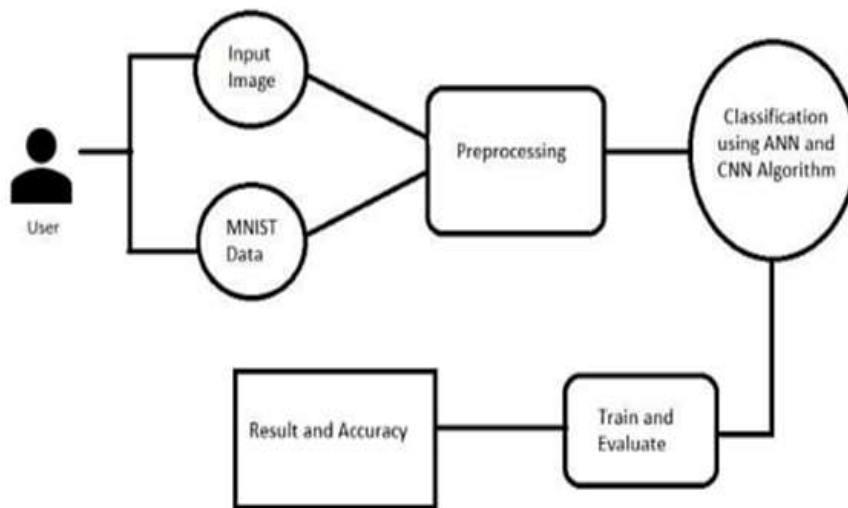
4.2 NON FUNCTIONAL REQUIREMENTS

FR No.	Non-Functional Requirement	Description
NFR-1	Usability	One of the very significant problems in pattern recognition applications is the recognition of handwritten characters. Applications for digit recognition include filling out forms, processing bankcheques, and sorting mail.
NFR-2	Security	1) The system generates a thorough description of the instantiation parameters, which might reveal information like the writing style, in addition to a categorization of the digit. 2) The generative models are capable of segmentation driven by recognition. 3) The procedure uses a relatively.
NFR-3	Reliability	The samples are used by the neural network to automatically deduce rules for reading handwritten digits. Furthermore, the network may learn more about handwriting and hence enhance its accuracy by increasing the quantity of training instances. Numerous techniques and algorithms, such as Deep Learning/CNN, SVM, Gaussian Naive Bayes, KNN, Decision Trees, Random Forests, etc., can be used to recognise handwritten numbers
NFR-4	Performance	Information is restricted to each user's limited access
NFR-5	Availability	Applications for digit recognition include filling out forms, processing bankchecks, and sorting mail.
NFR-6	Scalability	The system should be able to handle 1000 users accessing the website at the same time

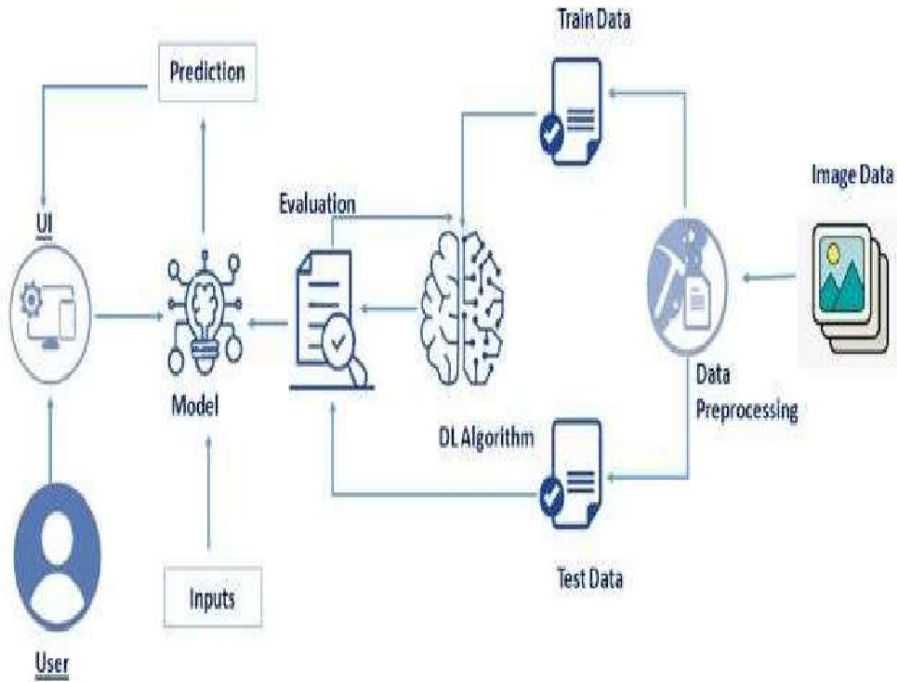
CHAPTER 5

PROJECT DESIGN

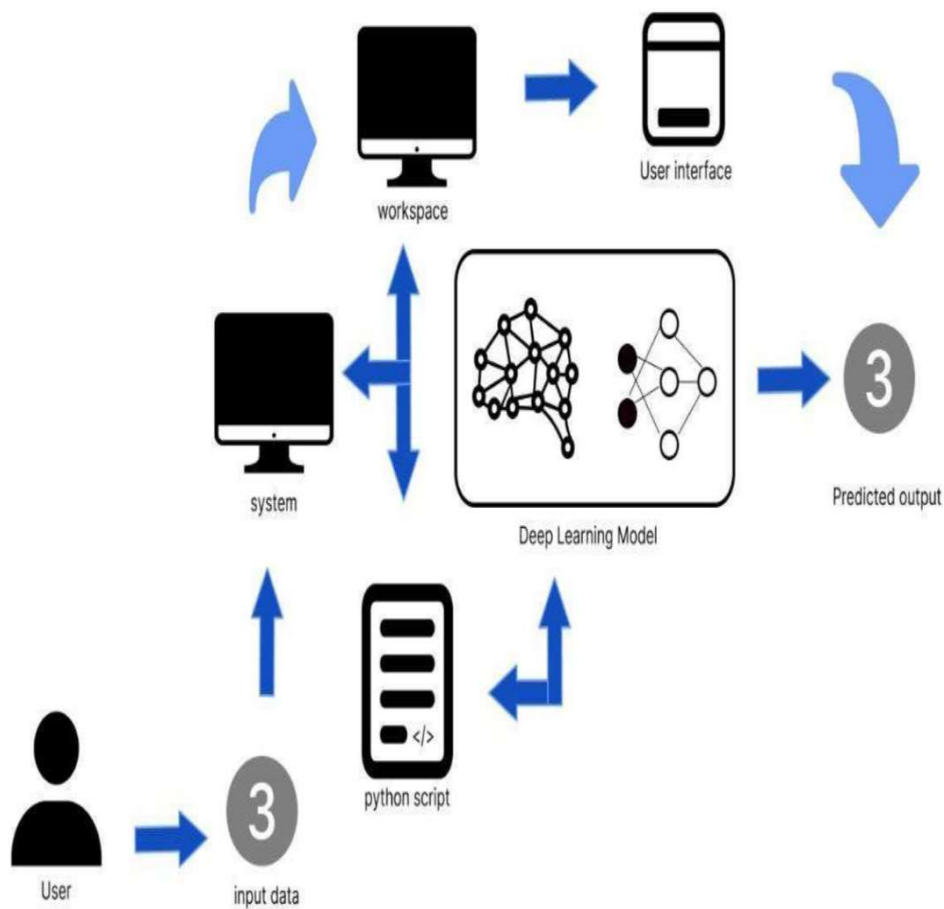
5.1 DATA FLOW DIAGRAM



5.2 SOLUTION & TECHNICAL ARCHITECTURE



Technology Architecture



Process Flow Diagram

5.3 USER STORIES

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer (Mobile user)	Home	USN-1	I can view the user manual and am aware of how to use this application.	I can see the limitation of this software & the awareness of how to use it.	Low	Sprint-1
		USN-2	I'm authorized to watch the instructional film to learn how to utilize this application's interface as a user.	I can learn how to use this application through a hands-on approach.	Low	Sprint-1
		USN-3	I am able to understand the directions for using this application as a user.	I am able to read the directions and utilize it.	Low	Sprint-2
	Recognize	USN-4	I get to select the image on this prediction page as a user.	I am able to select an image from our local system.	High	Sprint-2
	Predict	USN-5	I am permitted to upload and pick the image that will be submitted as a user.	From the system storage as well as any virtual storage, I may upload and select an image.	Medium	Sprint-3
		USN-6	I will train and evaluate the input as a user to ensure the output is as accurate as possible.	I am able to test and train the application.	High	Sprint-4
		USN-7	I have access to the MNIST data collection as a user.	To produce the precise output, I may access it.	Medium	Sprint-3

CHAPTER 6

PROJECT PLANNING AND SCHEDULING

6.1 SPRINT PLANNING AND ESTIMATION

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Data Collection	USN-1	As a user, I can collect the dataset from various resources with different handwritings.	10	Low	Sabitha J Nesika P
Sprint-1	Data Preprocessing	USN-2	As a user, I can load the dataset, handling the missing data, scaling and split data into train and test.	10	Medium	Nishanthini S Sangeetha M
Sprint-2	Model Building	USN-3	As a user, I will get an application with ML model which provides high accuracy of recognized handwritten digit.	5	High	Sabitha J Nesika P Nishanthini S Sangeetha M
Sprint-2	Add CNN layers	USN-4	Creating the model and adding the input, hidden, and output layers to it.	5	High	Nishanthini S Sabitha J Nesika P Sangeetha M
Sprint-2	Compiling the model	USN-5	With both the training data defined and model defined, it's time to configure the learning process.	2	Medium	Nesika P Sangeetha M Nishanthini S

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-2	Train & test the model	USN-6	As a user, let us train our model with our image dataset.	6	Medium	Sabitha J Nishanthini S Nesika P
Sprint-2	Save the model	USN-7	As a user, the model is saved & integrated with an android application or web application in order to predict something.	2	Low	Sangeetha M
Sprint-3	Building UI Application	USN-8	As a user, I will upload the handwritten digit image to the application by clicking a upload button.	5	High	Sabitha J Nesika P Sangeetha M
Sprint-3		USN-9	As a user, I can know the details of the fundamental usage of the application.	5	Low	Nishanthini S Sabitha J
Sprint-3		USN-10	As a user, I can see the predicted / recognized digits in the application.	5	Medium	Nishanthini S Sangeetha M
Sprint-4	Train the model on IBM	USN-11	As a user, I train the model on IBM and integrate flask/Django with scoring end point.	10	High	Nesika P Sabitha J Nishanthini S Sangeetha M
Sprint-4	Cloud Deployment	USN-12	As a user, I can access the web application and make the use of the product from anywhere.	10	High	Sabitha J Nesika P

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

CHAPTER 7

CODING & SOLUTIONING

```
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 computation function
from tensorflow.keras.layers import Dense, Flatten #Dense-Dense Layer is the regular deeply connected r
#flatten -used for flattening the input or change the dimension
from tensorflow.keras.layers import Conv2D #convolutional Layer
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 () #splitting the mnist data into train and test
```

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz>
 11490434/11490434 [=====] - 0s 0us/step

```
print (x_train.shape) #shape is used for give the dimension values #60000-rows 28x28-pixels
print (x_test.shape)
```

```
(60000, 28, 28)
(10000, 28, 28)
```

CHAPTER 8

TESTING

8.1 TEST CASES

Test case ID	Feature Type	Component	Test Scenario	Expected Result	Actual Result	Status
HP_TC_001	UI	Home Page	Verify UI elements in the Home Page	The Home page must be displayed properly	Working as expected	PASS
BE_TC_001	Functional	Backend	Check if all the routes are working properly	All the routes should properly work	Working as expected	PASS
RP_TC_001	UI	Result Page	Verify UI elements in the Result Page	The Result page must be 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
Total	5	1	3	3	12

8.2.2 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 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.

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 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 computation function
from tensorflow.keras.layers import Dense, Flatten #Dense-Dense Layer is the regular deeply connected r
#flatten -used for flattening the input or change the dimension
from tensorflow.keras.layers import Conv2D #convolutional Layer
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 () #splitting the mnist data into train and test
```

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz>
 11490434/11490434 [=====] - 0s 0us/step

```
print (x_train.shape) #shape is used for give the dimension values #60000-rows 28x28-pixels
print (x_test.shape)
```

```
(60000, 28, 28)
(10000, 28, 28)
```

```
#Reshaping to format which CNN expects (batch, height, width, channels)
x_train=x_train.reshape (60000, 28, 28, 1).astype('float32')
x_test=x_test.reshape (10000, 28, 28, 1).astype ('float32')
```

HOME PAGE (HTML)

```
8  <style>
9    body{
10      background-image: url('static/images/index6.jpg');
11      background-repeat: no-repeat;
12      background-size: cover;
13    }
14
15    #rectangle{
16      width:400px;
17      height:150px;
18      background-color: #5796a5;
19      border-radius: 25px;
20      position:absolute;
21      top:25%;
22      left:50%;
23      transform:translate(-50%,-50%);
24    }
25
26    #ans{
27      text-align: center;
28      font-size: 40px;
29      margin: 0 auto;
30      padding: 3% 5%;
31      padding-top: 15%;
32      color: white;
33    }
34
35  </style>
36  <body>
37    <div id="rectangle">
38      <h1 id="ans">Predicted Number : {{num}}</h1>
39    </div>
40  </body>
41  </html>
```

HOME PAGE (CSS)

```
1  #clear_button{
2    margin-left: 15px;
3    font-weight: bold;
4    color: blue;
5  }
6
7  #confidence{
8    font-family: 'Josefin Sans', sans-serif;
9    margin-top: 7.5%;
10 }
11
12 #content{
13   margin: 0 auto;
14   padding: 2% 15%;
15   padding-bottom: 0;
16 }
17
18 .welcome{
19   text-align: center;
20   position: relative;
21   color: honeydew;
22   background-color: greenyellow;
23   padding-top: 1%;
24   padding-bottom: 1%;
25   font-weight: bold;
26   font-family: 'Prompt', sans-serif;
27 }
28
29 #team_id{
30   text-align: right;
31   font-size: 25px;
32   padding-right: 3%;
33 }
34
35 #predict_button{
```

```
36   margin-right: 15px;
37   color: blue;
38   font-weight: bold;
39 }
40
41 #prediction_heading{
42   font-family: 'Josefin Sans', sans-serif;
43   margin-top: 7.5%;
44 }
45
46 #result{
47   font-size: 5rem;
48 }
49
50 #title{
51   padding: 1.5% 15%;
52   margin: 0 auto;
53   text-align: center;
54 }
55
56 .btn {
57   font-size: 15px;
58   padding: 10px;
59   -webkit-appearance: none;
60   background: #eee;
61   border: 1px solid #888;
62   margin-top: 20px;
63   margin-bottom: 20px;
64 }
65
66 .buttons_div{
67   margin-bottom: 30px;
68   margin-right: 80px;
69 }
```

```
71 .heading{
72   font-family: 'Varela Round', sans-serif;
73   font-weight: 700;
74   font-size: 2rem;
75   display: inline;
76 }
77
78 .leftside{
79   text-align: center;
80   margin: 0 auto;
81   margin-top: 2%;
82   /* padding-left: 10%; */
83 }
84
85 #frame{
86   margin-right: 10%;
87 }
88
89 .predicted_answer{
90   text-align: center;
91   margin: 0 auto;
92   padding: 3% 5%;
93   padding-top: 0;
94   /* padding-left: 10%; */
95 }
96
97 p{
98   font-family: 'Source Code Pro', monospace,sans-serif;
99   margin-top: 1%;
100 }
101
102 @media (min-width: 720px) {
103   .leftside{
104     padding-left: 10%;
105   }
```

app.py

```
13
14 app = Flask(__name__)
15 app.config['UPLOAD_FOLDER'] = UPLOAD_FOLDER
16
17 model = load_model("mnistCNN.h5")
18
19
20 @app.route('/')
21 def index():
22     return render_template('index.html')
23
24
25 @app.route('/predict', methods=['GET', 'POST'])
26 def upload():
27     if request.method == "POST":
28         f = request.files["image"]
29         filepath = secure_filename(f.filename)
30         f.save(os.path.join(app.config['UPLOAD_FOLDER'], filepath))
31
32         upload_img = os.path.join(UPLOAD_FOLDER, filepath)
33         img = Image.open(upload_img).convert("L") # convert image to monochrome
34         img = img.resize((28, 28)) # resizing of input image
35
36         im2arr = np.array(img) # converting to image
37         im2arr = im2arr.reshape(1, 28, 28, 1) # reshaping according to our requirement
38
39         pred = model.predict(im2arr)
40
41         num = np.argmax(pred, axis=1) # printing our labels
42
43         return render_template('predict.html', num=str(num[0]))
44
45
46 if __name__ == '__main__':
47     app.run(debug=True, threaded=False)
```


index.html

```

3 <head>
4 <title>Digit Recognition WebApp</title>
5
6 <meta name="viewport" content="width=device-width">
7 <!-- GoogleFont -->
8 <link href="https://fonts.googleapis.com/css2?family=Prompt:wght@600&display=swap" rel="stylesheet">
9 <link href="https://fonts.googleapis.com/css2?family=Varela+Round&display=swap" rel="stylesheet">
10 <link href="https://fonts.googleapis.com/css2?family=Source+Code+Pro:wght@500&display=swap" rel="stylesheet">
11 <link href="https://fonts.googleapis.com/css2?family=Calistoga|Josefin+Sans:400,700|Pacifico&display=swap" rel="stylesheet">
12 <!-- bootstrap -->
13 <link rel="stylesheet" href="https://stackpath.bootstrapcdn.com/bootstrap/4.3.1/css/bootstrap.min.css" integrity="sha384-ggOyR0iXCbMQv3Xipma34ND+dH/1fQ784/j6cY/iJTQUOhcWr7x9"
14 <link rel="stylesheet" type="text/css" href= "{{ url_for('static', filename='css/style.css') }}">
15 <!-- fontawesome -->
16 <script src="https://kit.fontawesome.com/b3aed9cb07.js" crossorigin="anonymous"></script>
17
18 <script src="https://code.jquery.com/jquery-3.3.1.slim.min.js" integrity="sha384-q8i/X+965Dz00rT7abK41JStQIAqVgRVzpbzo5smXKp4YfRvH+8abTTE1P6jizo" crossorigin="anonymous"></
19 <script src="https://cdnjs.cloudflare.com/ajax/libs/popper.js/1.14.7/umd/popper.min.js" integrity="sha384-U02eT0CpHqdSJQ6hJty5KVphtPhzWj9WO1clHTMGa3JDZwrnQq4sF86dIHNDz0W1" c
20 <script src="https://stackpath.bootstrapcdn.com/bootstrap/4.3.1/js/bootstrap.min.js" integrity="sha384-JjSmVgyd0p3pX81rRibZUAYoIIy60rQ6VrjIEaFf/njGzIxFSf4x0xIM+h007jRM" cros
21 <script src="https://cdn.jsdelivr.net/npm/@tensorflow/tfjs@latest"></script>
22
23 </head>
24
25 <script>
26 function preview() {
27   frame.src=URL.createObjectURL(event.target.files[0]);
28 }
29
30 $(document).ready(function() {
31   $('#clear_button').on('click', function() {
32     $('#image').val('');
33     $('#frame').attr('src','');
34   });
35 });
36

```

```

41 <h1 class="welcome">IBM PROJECT
42 <div id="team_id">TEAM ID : PNT2022TMID05025</div>
43 </h1>
44 <section id="title">
45 <h4 class="heading">Handwritten Digit Recognition Website</h4>
46 <br><br>
47 <p>
48 The website is designed to predict the handwritten digit.
49 </p>
50 <p>
51 Handwriting recognition is one of the compelling research works going on because every individual in this world
52 has their own style of writing. It is the capability of the computer to identify and understand
53 handwritten digits or characters automatically. Because of the progress in the field of science and technology,
54 everything is being digitalized to reduce human effort.</p>
55
56 <br>
57 <p> Hence, there comes a need for handwritten digit recognition in many real-time applications.
58 MNIST data set is widely used for this recognition process and it has 70000 handwritten digits.
59 We use Artificial neural networks to train these images and build a deep learning model.
60 Web application is created where the user can upload an image of a handwritten digit.
61 This image is analyzed by the model and the detected result is returned on to UI</p>
62
63 </section>
64
65 <section id="content">
66
67 <div class="leftside">
68 <form action="/predict" method="POST" enctype="multipart/form-data">
69 <label>Select a image:</label>
70 <input id="image" type="file" name="image" accept="image/png, image/jpeg" onchange="preview()"><br><br>
71 <img id="frame" src="" width="100px" height="100px"/>
72 <div class="buttons_div">
73 <button type="submit" class="btn btn-dark" id="predict_button">Predict</button>
74 <button type="button" class="btn btn-dark" id="clear_button">&nbsp;Clear &nbsp;</button>
75 </div>

```

GITHUB

<https://github.com/IBM-EPBL/IBM-Project-8922-1658938041>

PROJECT DEMO LINK

https://drive.google.com/file/d/1b8egZ_U12YF5csdygfOsa56kmNLFX5NV/view?usp=sharing

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