



# MNIST HANDWRITTEN DIGIT RECOGNITION

*SQE Project*



# GROUP MEMBERS

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# INTRODUCTION

- ❑ Handwritten digits recognition could be a well-researched subarea.
- ❑ Involved with learning models to differentiate pre-segmented written digits.
- ❑ Main application of machine learning strategies has determined efficacious in conformist decisive systems.
- ❑ In 21<sup>st</sup> century written digit communication has its own normal
- ❑ Challenges in handwritten characters recognition lies within the variation and distortion of handwritten.
- ❑ The challenges are principally caused by the big variation of individual writing styles.
- ❑ Strong feature extraction is extremely important.
- ❑ Handwritten digit recognition has obtained heap of concentration.
- ❑ Character recognition system may function as a cornerstone.



# WHICH ALGORITHM IS USED AND WHY?

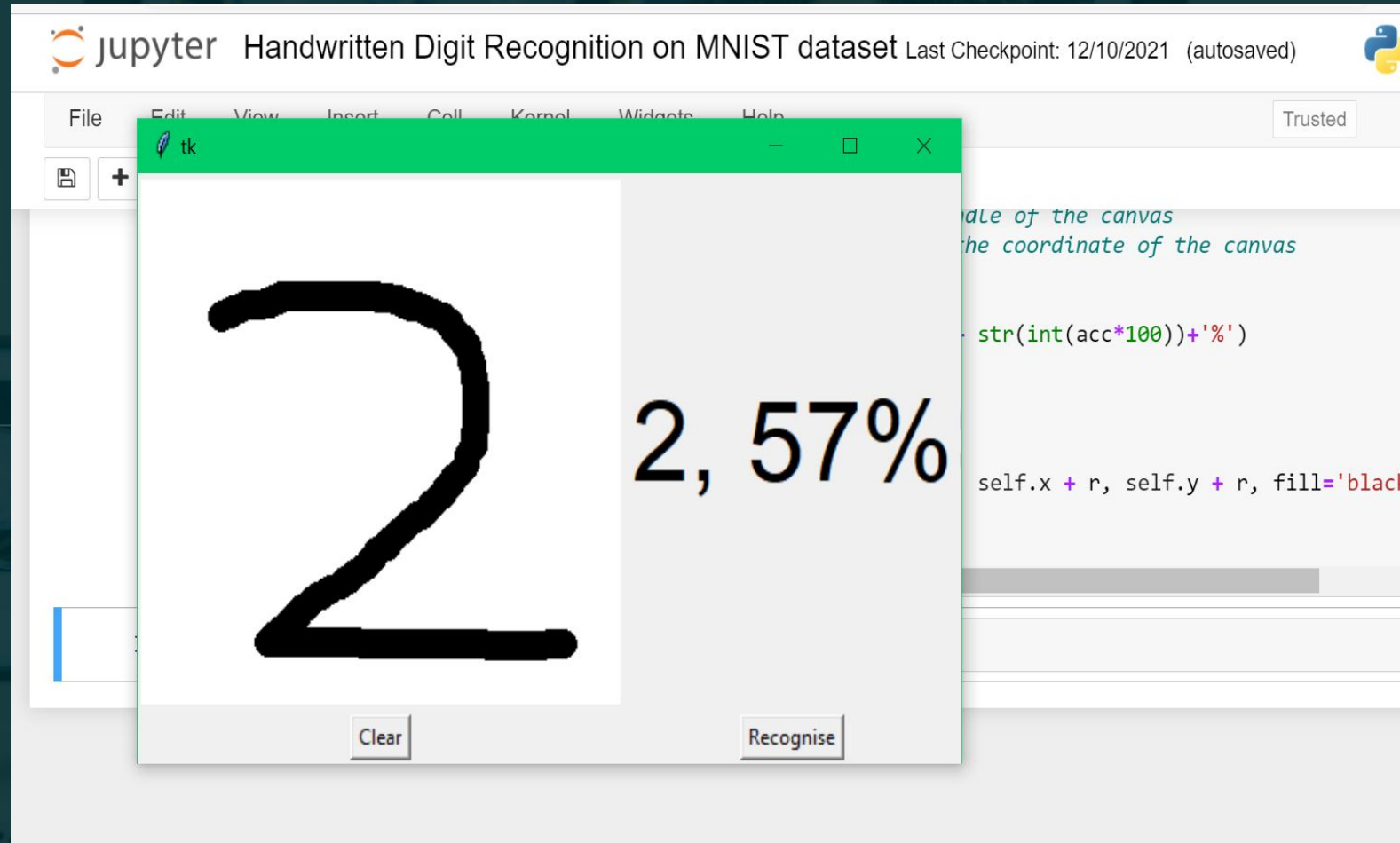
- ❑ Neural Network: Training and testing was done.
- ❑ CNN was found to have highest accuracy of 98%.
- ❑ It was used for the model.
- ❑ Both were tested for results in GUI user-based application.

# DRY RUN OF ALGORITHM

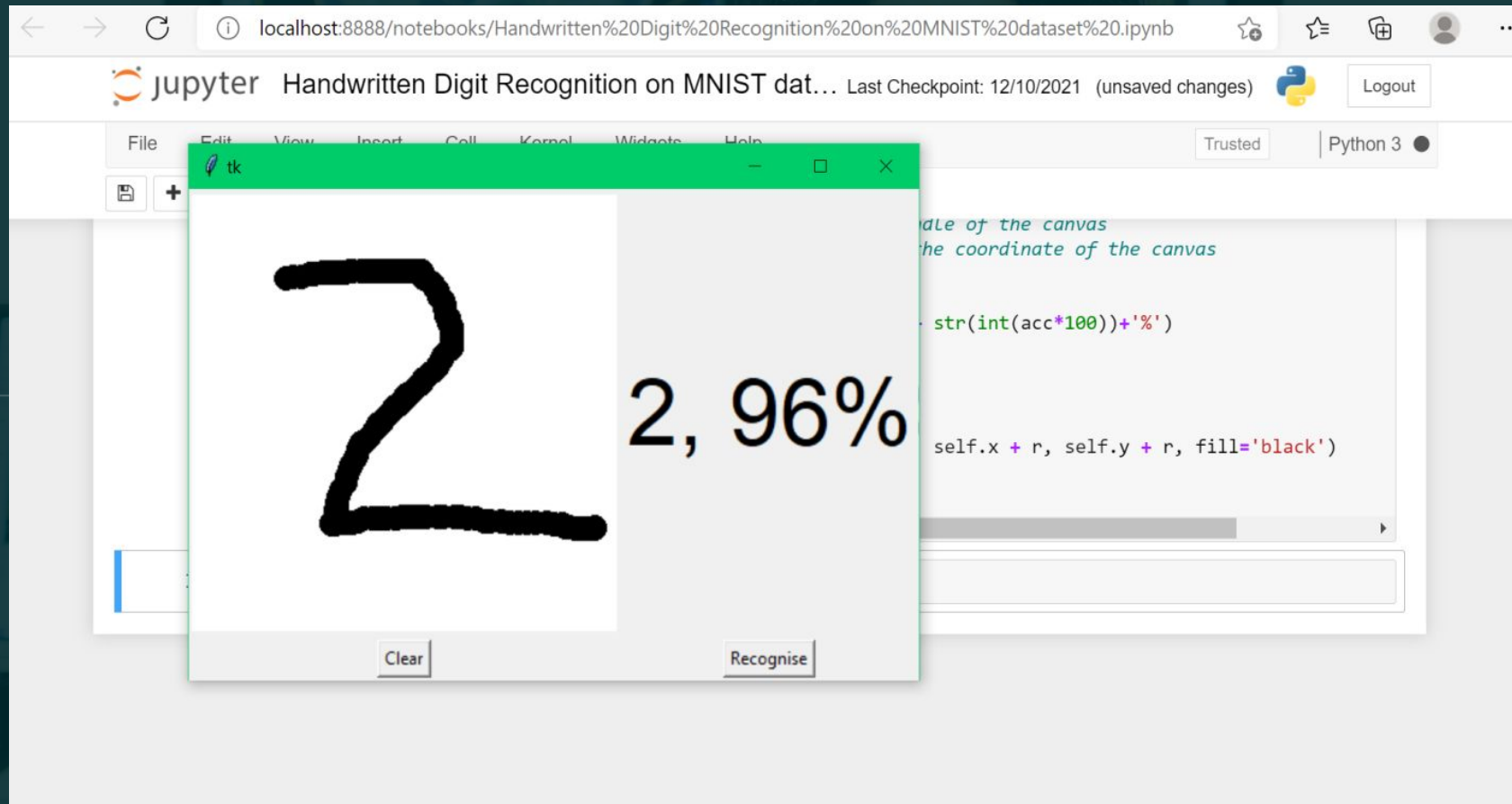
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- ❑ Import the libraries and load the MNIST dataset
- ❑ Data Preprocess and Normalize
- ❑ Create the model
- ❑ Train the model
- ❑ Evaluate the model
- ❑ Create GUI to predict digits

# DRY RUN OF ALGORITHM



# DRY RUN OF ALGORITHM





# TOOLS AND TECHNOLOGIES USED:

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- ❑ We will be using a special type of deep neural network that is **Convolutional Neural Networks**. In the end, we are going to build a GUI in which you can draw the digit and recognize it straight away/
- ❑ To achieve high performance on the handwritten digit recognition task using the **MNIST** dataset and build a **GUI App** based on **Tkinter**.



# OUTPUT

## Recognition Through Fully Covolutional Network Model.

```
In [6]: 1 #Building FCN Model
        2 ###4.Build the model
        3 model = Sequential()
        4 model.add(Dense(hidden1, input_dim=input_size, activation='relu'))
        5 # output = relu (dot (W, input) + bias)
        6 model.add(Dense(hidden2, activation='relu'))
        7 model.add(Dense(classes, activation='softmax'))
        8
        9 # Compilation
       10 model.compile(loss='categorical_crossentropy',
       11               metrics=['accuracy'], optimizer='sgd')
       12 model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 400)	314000
dense_1 (Dense)	(None, 20)	8020
dense_2 (Dense)	(None, 10)	210
Total params: 322,230		
Trainable params: 322,230		
Non-trainable params: 0		

```
In [7]: 1 #Training the model
        2 # Fitting on Data
        3 model.fit(X_train, Y_train, batch_size=batch_size, epochs=10, verbose=2)
        4 ###5.Test
```

```
Epoch 1/10
300/300 - 5s - loss: 1.5074 - accuracy: 0.6038
Epoch 2/10
300/300 - 2s - loss: 0.7047 - accuracy: 0.8234
Epoch 3/10
300/300 - 2s - loss: 0.5072 - accuracy: 0.8679
Epoch 4/10
300/300 - 2s - loss: 0.4235 - accuracy: 0.8858
Epoch 5/10
300/300 - 2s - loss: 0.3772 - accuracy: 0.8948
Epoch 6/10
300/300 - 2s - loss: 0.3470 - accuracy: 0.9022
Epoch 7/10
300/300 - 2s - loss: 0.3251 - accuracy: 0.9081
Epoch 8/10
300/300 - 2s - loss: 0.3076 - accuracy: 0.9123
Epoch 9/10
300/300 - 2s - loss: 0.2931 - accuracy: 0.9164
Epoch 10/10
300/300 - 2s - loss: 0.2807 - accuracy: 0.9197
```

Out[7]: <keras.callbacks.History at 0x22592ec2e50>

# OUTPUT

```
In [8]: 1 #Testing the model
2 score = model.evaluate(X_test, Y_test, verbose=1)
3 print('\n' 'Test accuracy:', score[1])
4 mask = range(10,20)
5 X_valid = X_test[mask]
6 y_pred = (model.predict(X_valid) > 0.5).astype("int32")
7 print(y_pred)
8 plt.figure(figsize=(20, 4))
9 for i in range(n):
10     # display original
11     ax = plt.subplot(2, n, i + 1)
12     plt.imshow(X_valid[i].reshape(28, 28))
13     plt.gray()
14     ax.get_xaxis().set_visible(False)
15     ax.get_yaxis().set_visible(False)
16 plt.show()
17 plt.close()
18 model.save('C:\\Users\\Admin\\Downloads\\FCNN')
```

313/313 [-----] - 1s 3ms/step - loss: 0.2656 - accuracy: 0.9228

Test accuracy: 0.922800044822693

```
[[1 0 0 0 0 0 0 0 0 0]
 [0 0 0 0 0 0 1 0 0 0]
 [0 0 0 0 0 0 0 0 0 1]
 [1 0 0 0 0 0 0 0 0 0]
 [0 1 0 0 0 0 0 0 0 0]
 [0 0 0 0 0 1 0 0 0 0]
 [0 0 0 0 0 0 0 0 0 1]
 [0 0 0 0 0 0 0 1 0 0]
 [0 0 0 1 0 0 0 0 0 0]
 [0 0 0 0 1 0 0 0 0 0]]
```

0 6 9 0 1 5 9 7 8 4

INFO:tensorflow:Assets written to: C:\Users\Admin\Downloads\FCNN\assets

# OUTPUT

## Testing models through GUI Application

```
In [17]: 1 #GUI
2 from keras.models import load_model
3 from tkinter import *
4 import tkinter as tk
5 import win32gui
6 from PIL import ImageGrab, Image
7 import numpy as np
8 model = load_model('C:\\\\Users\\Admin\\Downloads\\CNN')
9 def predict_digit(img):
10     #resize image to 28x28 pixels
11     img = img.resize((28,28))
12     #convert rgb to grayscale
13     img = img.convert('L')
14     img = np.array(img)
15     #reshaping to support our model input and normalizing
16     img = img.reshape(1,28,28,1)
17     img = img/255.0
18     #predicting the class
19     res = model.predict([img])[0]
20     return np.argmax(res), max(res)
21 class App(tk.Tk):
22     def __init__(self):
23         tk.Tk.__init__(self)
24         self.x = self.y = 0
25         # Creating elements
26         self.canvas = tk.Canvas(self, width=300, height=300, bg = "white", cursor="cross")
27         self.label = tk.Label(self, text="Thinking..", font=("Helvetica", 48))
28         self.classify_btn = tk.Button(self, text = "Recognise", command = self.classify_handwriting)
29         self.button_clear = tk.Button(self, text = "Clear", command = self.clear_all)
30         # Grid structure
31         self.canvas.grid(row=0, column=0, pady=2, sticky=W, )
32         self.label.grid(row=0, column=1, pady=2, padx=2)
33         self.classify_btn.grid(row=1, column=1, pady=2, padx=2)
34         self.button_clear.grid(row=1, column=0, pady=2)
35         #self.canvas.bind("<Motion>", self.start_pos)
36         self.canvas.bind("<B1-Motion>", self.draw_lines)
37     def clear_all(self):
38         self.canvas.delete("all")
39     def classify_handwriting(self):
40         HWND = self.canvas.winfo_id() # get the handle of the canvas
41         rect = win32gui.GetWindowRect(HWND) # get the coordinate of the canvas
42         im = ImageGrab.grab(rect)
43         digit, acc = predict_digit(im)
44         self.label.configure(text= str(digit)+'', '+ str(int(acc*100))+'%')
45     def draw_lines(self, event):
46         self.x = event.x
47         self.y = event.y
48         r=8
49         self.canvas.create_oval(self.x-r, self.y-r, self.x + r, self.y + r, fill='black')
50 app = App()
51 mainloop()
```



THANK  
YOU