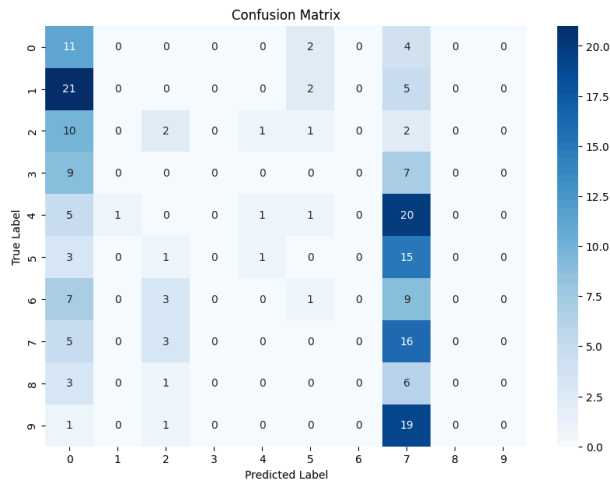
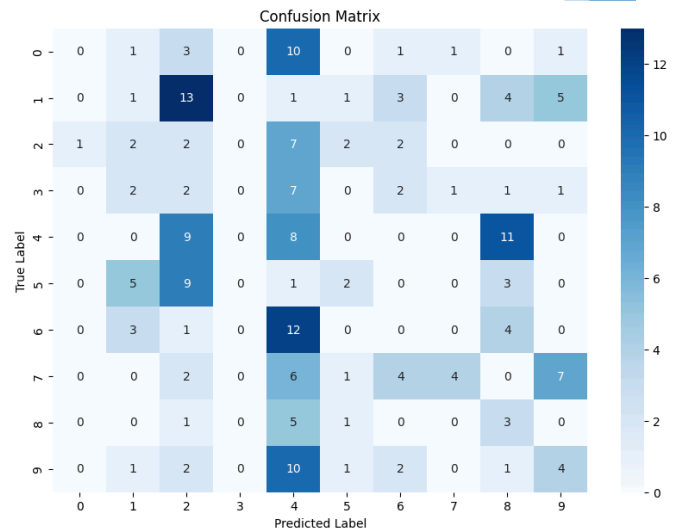


Put Your Code Results Here:

First test



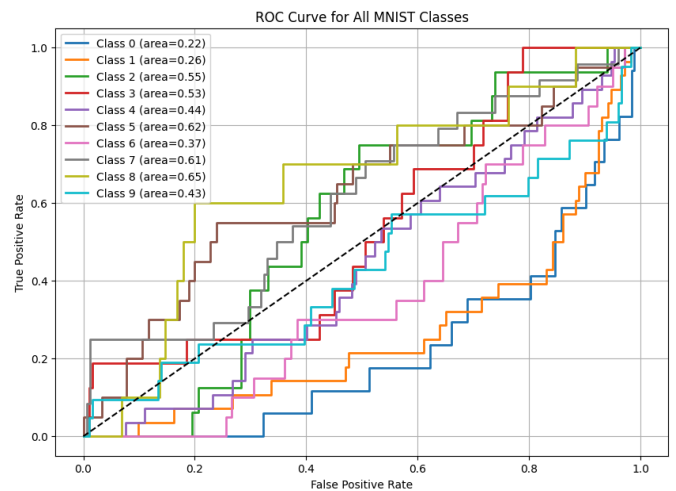
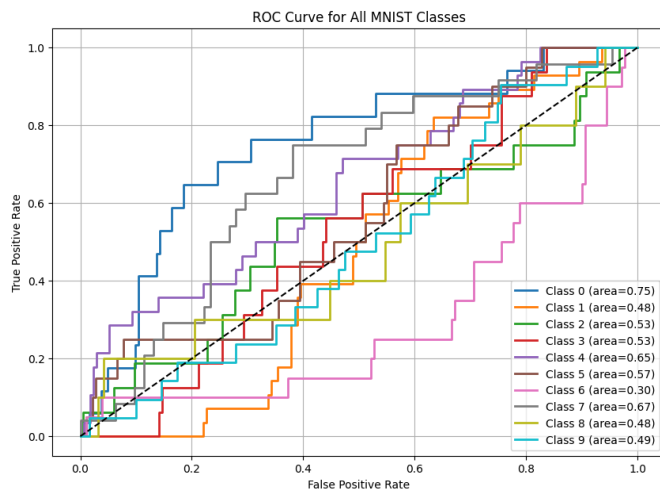
Second time test



The number 1 is easily predicted to be a 0, possibly because certain characteristics of handwritten digits 1 and 0 are easily confused.

The number 7 was also more incorrectly judged as other numbers.

The number 4 is an obvious center of confusion, with many numbers easily mistaken for four. The numbers 7 and 9 are also somewhat confused with each other.



input_size = 784
hidden1 = 64
hidden2 = 32
output_size = 10

Use **ReLU** For activation Function

input_size = 784
hidden1 = 128
hidden2 = 64
hidden3 = 32
output_size = 10

Use **tanh** For activation Function



| | precision | recall | f1-score | support | | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|--------------|-----------|--------|----------|---------|
| 0 | 0.1467 | 0.6471 | 0.2391 | 17 | 0 | 0.0000 | 0.0000 | 0.0000 | 17 |
| 1 | 0.0000 | 0.0000 | 0.0000 | 28 | 1 | 0.0667 | 0.0357 | 0.0465 | 28 |
| 2 | 0.1818 | 0.1250 | 0.1481 | 16 | 2 | 0.0455 | 0.1250 | 0.0667 | 16 |
| 3 | 0.0000 | 0.0000 | 0.0000 | 16 | 3 | 0.0000 | 0.0000 | 0.0000 | 16 |
| 4 | 0.3333 | 0.0357 | 0.0645 | 28 | 4 | 0.1194 | 0.2857 | 0.1684 | 28 |
| 5 | 0.0000 | 0.0000 | 0.0000 | 20 | 5 | 0.2500 | 0.1000 | 0.1429 | 20 |
| 6 | 0.0000 | 0.0000 | 0.0000 | 20 | 6 | 0.0000 | 0.0000 | 0.0000 | 20 |
| 7 | 0.1553 | 0.6667 | 0.2520 | 24 | 7 | 0.6667 | 0.1667 | 0.2667 | 24 |
| 8 | 0.0000 | 0.0000 | 0.0000 | 10 | 8 | 0.1111 | 0.3000 | 0.1622 | 10 |
| 9 | 0.0000 | 0.0000 | 0.0000 | 21 | 9 | 0.2222 | 0.1905 | 0.2051 | 21 |
| accuracy | | | 0.1500 | 200 | accuracy | | | 0.1200 | 200 |
| macro avg | 0.0817 | 0.1474 | 0.0704 | 200 | macro avg | 0.1482 | 0.1204 | 0.1058 | 200 |
| weighted avg | 0.0923 | 0.1500 | 0.0714 | 200 | weighted avg | 0.1636 | 0.1200 | 0.1114 | 200 |

Adding one more layer (32 neuro units) makes the neural network deeper. And change activation function from ReLU to tanh but look like doesn't make ACC value higher

1. Explain how you designed your model (number of layers, neurons, and activation functions).

What changes did you make to improve the accuracy, and how did those changes affect the results?

*Please attach the performance results before and after your improvements.

At first, I made my model with a pretty simple structure. It had an input layer with 784 neurons, then two hidden layers with 64 and 32 neurons, and an output layer with 10 neurons. For the activation functions, I used ReLU in the hidden layers and Softmax in the output layer.

To make the accuracy better, I tried some changes. I increased the number of neurons in the hidden layers to 128 and 64, and I even added one more hidden layer with 32 neurons. I also changed the activation function from ReLU to tanh because I thought it might help the model find more complex patterns.

But the changes didn't work as I hoped. Surprisingly, the accuracy went down from 15% to 12%. I think this happened because the model wasn't trained with backpropagation and was just using random weights. Adding more layers just made the model more complicated without actually learning anything useful. The extra layers probably just added more noise instead of helping the model recognize the digits better.

2. Based on your evaluation results (confusion matrix, ROC, etc.), how well did the model perform? Which classes are harder to predict? Why do you think that happened?

Honestly, the model didn't do well at all. The accuracy was really low, only around 12–15%, which is just a little bit better than random guessing. The confusion matrix shows that the model had a hard time telling apart most of the numbers. For example, it often got confused with 3, 6, 8, and 9—they were almost always guessed wrong, and the precision and recall were close to zero.

The main problem is that the model didn't really learn how to recognize the numbers because it was just using random weights without training. That's why it mostly guessed instead of making accurate predictions. Also, some numbers like 1 and 7 were often mistaken for 0. I think this is because handwritten 1 and 7 can



look a bit like 0 if written in a certain way, which makes the model confused.

One thing that stood out is that the numbers 0 and 9 were recognized a bit better than the others. I think it's because 0 has a unique round shape, and 9 often has a distinct tail, so they are a bit easier to tell apart even when using random weights.

Overall, just adding more layers without training didn't help—in fact, it made things worse. The model needs proper training with backpropagation to actually learn the features instead of just guessing randomly.

