

NAME - JATAN SAHU  
ID-202218061  
LAB - 04  
SUBJECT - DEEP LEARNING

## ▼ INSTRUCTIONS

### 1.Introduction

Working with images This lab aims to demonstrate the efficacy of ANNs in image processing, alongside regression analysis on the provided dataset. Additionally, it provides an opportunity to gain expertise in implementing normalization techniques and integrating skip connections within a deep learning model.

### 2.Working with SKIP connections and regularization

Follow the given notebook for building the models with skip connection and regularization.

### 3.Dataset

- 1.MNIST: The stepping stone or Hello world of Deep Learning this dataset contains images of handwritten digits
- 2.CIFAR-10: Contains colored images of various objects
- 3.Auto-mpg: The data is technical spec of cars. In this regression dataset we need to predict 'mpg' attribute from other column values.

### 4.Tasks:

#### 4.1 Classification

- 1.Load and visualize the images from the dataset.
- 2.Apply preprocessing and encoding to labels.
- 3.Define the ANN model for image classification. Include normalization and skip connections in our model.
- 4.Experiment with the different activation functions and loss functions while training the model
- 5.Analyze ANN model performance with different batch sizes (test of 3 different batch size) and learning rates (3 different learning rates)
- 6.Plot the accuracy for train and test data
- 7.You can use matplotlib plots to present your analysis for different hyper- parameters
- 8.Evaluate the model's performance using different performance matrices discussed in the class

#### 4.2 Regression

- 1.Load and preprocess the given data.
- 2.Build ANN model with regularization and skip connections and train it on the given data.
- 3.Analyze ANN model performance with different batch sizes (test of 3 different batch size) and learning rates (3 different learning rates).
- 4.Plot mse, mae and rmse for different batch size and learning rates.

## Classification TASK

### DATASET01 - CIFAR DATA

## ▼ 1. Importing Libraries

```
1 from tensorflow.keras.models import Model
2 from tensorflow.keras.layers import Dense, Input, Dropout, BatchNormalization, Add, Concatenate
3 from keras.utils import to_categorical, plot_model
```

```

4 import numpy as np
5 import pandas as pd
6 import matplotlib.pyplot as plt
7 from sklearn.datasets import load_linnerud
8 from sklearn import datasets
9 from sklearn.model_selection import train_test_split
10 import numpy as np
11 from keras.datasets import cifar10
12 from keras.models import Sequential, Model
13 from keras.layers import Dense, BatchNormalization, Flatten, Input, Concatenate
14 from keras.optimizers import SGD
15 from keras.utils import to_categorical
16 import matplotlib.pyplot as plt
17 from sklearn.metrics import classification_report, confusion_matrix

```

## ▼ 2.Importing dataset using Tensorflow

```

1 import tensorflow as tf
2
3 # Load CIFAR-10 dataset
4 (x_train, y_train), (x_test, y_test) = tf.keras.datasets.cifar10.load_data()

Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz
170498071/170498071 [=====] - 11s 0us/step

```

Learning rate - 0.001

## ▼ 2. ML Training model PIPELINE

### 2.1Splitting data

### 2.2Preprocessing

### 2.3Flattening data

### 2.4Adding activation function

### 2.5Normalisation

### 2.6Skip Connectin

### 2.7 Adding Optimizer,loss function, metrics

## 3. Experiment with

### 3.1Activation functions

### 3.2Loss functions

### 3.3Batch sizes

### 3.4Learning rates

```

1 # Load the CIFAR-10 dataset and preprocess it
2 (x_train, y_train), (x_test, y_test) = cifar10.load_data()
3 x_train = x_train.astype('float32') / 255
4 x_test = x_test.astype('float32') / 255
5 y_train_encoded = to_categorical(y_train, num_classes=10)
6 y_test_encoded = to_categorical(y_test, num_classes=10)
7
8 # Define a function to create and train the model
9 def create_and_train_model(activation_func, loss_func, batch_size, learning_rate):
10     input_layer = Input(shape=(32, 32, 3))

```

```

11 x = Flatten()(input_layer)
12 x = Dense(128, activation=activation_func)(x)
13 x = BatchNormalization()(x)
14
15 skip_connection = x # Save a copy of the output for the skip connection
16
17 x = Dense(64, activation=activation_func)(x)
18 x = BatchNormalization()(x)
19
20 x = Dense(32, activation=activation_func)(x)
21 x = BatchNormalization()(x)
22
23 # Concatenate the output of the skip connection with the current output
24 x = Concatenate()([x, skip_connection])
25
26 output_layer = Dense(10, activation='softmax')(x)
27
28 model = Model(inputs=input_layer, outputs=output_layer)
29
30 optimizer = SGD(learning_rate=learning_rate)
31 model.compile(optimizer=optimizer, loss=loss_func, metrics=['accuracy'])
32
33 history = model.fit(x_train, y_train_encoded, validation_data=(x_test, y_test_encoded),
34                     batch_size=batch_size, epochs=10, verbose=0)
35
36 return history, model
37
38 # Experiment with different activation functions, loss functions, batch sizes, and learning rates
39 activation_functions = ['relu', 'tanh', 'sigmoid']
40 loss_functions = ['categorical_crossentropy', 'mean_squared_error']
41 batch_sizes = [32, 64, 128]
42 learning_rates = [0.001]
43
44 results = []
45
46 for activation_func in activation_functions:
47     for loss_func in loss_functions:
48         for batch_size in batch_sizes:
49             for learning_rate in learning_rates:
50                 history, model = create_and_train_model(activation_func, loss_func, batch_size, learning_rate)
51                 accuracy = history.history['accuracy'][-1]
52                 val_accuracy = history.history['val_accuracy'][-1]
53                 results.append((activation_func, loss_func, batch_size, learning_rate, accuracy, val_a

```

## ▼ 4.Plots

### 4.1Accuracy for train and test

## 5.Evaluation

### 5.1Classification Report

### 5.2Confusion Matrix

```

1 # Plot the accuracy for train and test data for each combination
2 for result in results:
3     activation_func, loss_func, batch_size, learning_rate, accuracy, val_accuracy, history, model = result
4     label = f'{activation_func}_{loss_func}_{batch_size}_{learning_rate}'
5
6     # Create a new figure and axes for each combination
7     plt.figure()
8     plt.plot(history.history['accuracy'], label=f'{label}_train', linestyle='-')
9     plt.plot(history.history['val_accuracy'], label=f'{label}_test', linestyle='--')
10    plt.xlabel('Epochs')
11    plt.ylabel('Accuracy')

```

```

12 plt.title(f'Model with {activation_func} activation, {loss_func} loss, batch size {batch_size}, le
13 plt.legend(loc='lower right')
14 plt.show()
15
16 # Evaluate the model's performance using different performance metrics
17 for result in results:
18     activation_func, loss_func, batch_size, learning_rate, accuracy, val_accuracy, history, model = re
19     y_pred = model.predict(x_test)
20     y_pred_classes = np.argmax(y_pred, axis=1)
21     y_true = y_test.squeeze()
22
23     print(f'Model with {activation_func} activation, {loss_func} loss, batch size {batch_size}, learni
24
25     # Classification Report
26     print(classification_report(y_true, y_pred_classes))
27
28     # Confusion Matrix
29     cm = confusion_matrix(y_true, y_pred_classes)
30     print("Confusion Matrix:")
31     print(cm)

```

## ▼ DATASET 02 -MNIST DATA

```
1 from tensorflow.keras.datasets import mnist
```

```
1 (train_images, train_labels), (test_images, test_labels) = mnist.load_data()
2 train_images.shape
```

```

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11490434/11490434 [=====] - 1s 0us/step
(60000, 28, 28)

```

```
1 train_labels.shape
```

```
(60000,)
```

```
1 test_images.shape
```

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

```
1 train_labels
```

```
array([5, 0, 4, ..., 5, 6, 8], dtype=uint8)
```

```
1 train_images
```

```

array([[0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0],
       ...,
       [0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0]],

      [[0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0],
       ...,
       [0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0]],

      [[0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0],
       ...,
       [0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0]],

      ...,

      [[0, 0, 0, ..., 0, 0, 0],
       [0, 0, 0, ..., 0, 0, 0],

```

```

[0, 0, 0, ..., 0, 0, 0],
...,
[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0]],

[[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0],
...,
[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0]],

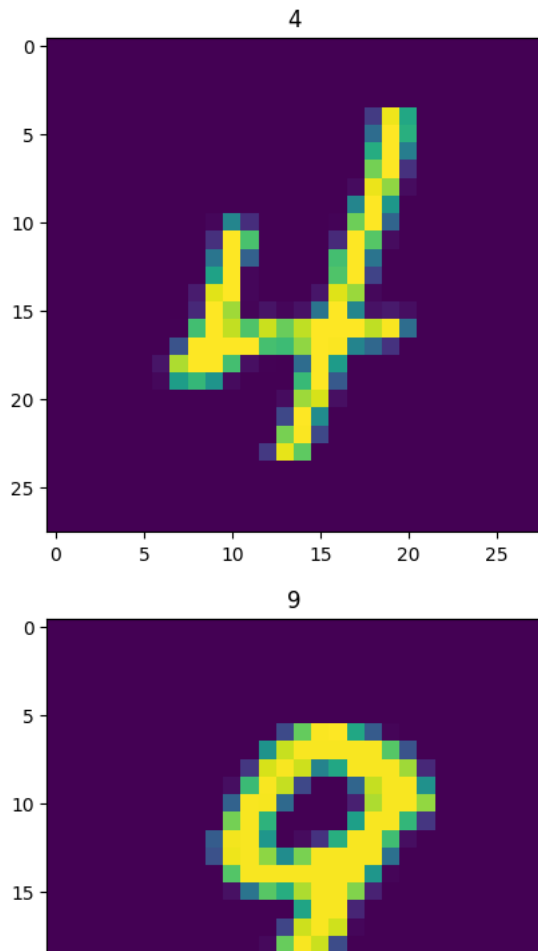
[[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0],
...,
[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0]], dtype=uint8)

```

```

1 # Plotting images
2 for index in np.random.randint(0,60000,4):
3     plt.imshow(train_images[index,:,:])
4     plt.title(train_labels[index])
5     plt.show()

```



▼ 2. Apply preprocessing and encoding to labels.

```
1 # Encoding Labels: Convert integer labels to one-hot encoded vectors
2 train_labels_encoded = to_categorical(train_labels)
3 test_labels_encoded = to_categorical(test_labels)
```

```
1 # Preprocessing: Normalize pixel values to the range [0, 1]
2 train_images = train_images.astype('float32') / 255
3 test_images = test_images.astype('float32') / 255
4
5 # Reshape the image data
6 train_images = train_images.reshape((train_images.shape[0], -1))
7 test_images = test_images.reshape((test_images.shape[0], -1))
```

```
1 train_labels_encoded.shape, test_labels_encoded.shape
```

```
((60000, 10), (10000, 10))
```

```
1 train_images.shape, train_images.shape
```

```
((60000, 784), (60000, 784))
```

▼ 3. Define the ANN model for image classification. Include normalization and skip connections in our model.

```
1 import numpy as np
2 from keras.datasets import mnist
3 from keras.models import Sequential, Model
4 from keras.layers import Dense, BatchNormalization, Flatten, Input, Concatenate
5 from keras.optimizers import SGD
6 from keras.utils import to_categorical
7 import matplotlib.pyplot as plt
8 from sklearn.metrics import classification_report, confusion_matrix
9
10 (train_images, train_labels), (test_images, test_labels) = mnist.load_data()
```

```
11 train_images = train_images.astype('float32') / 255
12 test_images = test_images.astype('float32') / 255
13 train_labels_encoded = to_categorical(train_labels)
14 test_labels_encoded = to_categorical(test_labels)
15
16 # Define a function to create and train the model
17 def create_and_train_model(activation_func, loss_func, batch_size, learning_rate):
18     input_layer = Input(shape=(28, 28))
19     x = Flatten()(input_layer)
20     x = Dense(128, activation=activation_func)(x)
21     x = BatchNormalization()(x)
22
23     skip_connection = x # Save a copy of the output for the skip connection
24
25     x = Dense(64, activation=activation_func)(x)
26     x = BatchNormalization()(x)
27
28     x = Dense(32, activation=activation_func)(x)
29     x = BatchNormalization()(x)
30
31     # Concatenate the output of the skip connection with the current output
32     x = Concatenate()([x, skip_connection])
33
34     output_layer = Dense(10, activation='softmax')(x)
35
36     model = Model(inputs=input_layer, outputs=output_layer)
37
38     optimizer = SGD(learning_rate=learning_rate)
39     model.compile(optimizer=optimizer, loss=loss_func, metrics=['accuracy'])
40
41     history = model.fit(train_images, train_labels_encoded, validation_data=(test_images, test_labels_
42         batch_size=batch_size, epochs=10, verbose=0)
43
44     return history, model
45
46 # Experiment with different activation functions, loss functions, batch sizes, and learning rates
47 activation_functions = ['relu', 'tanh', 'sigmoid']
48 loss_functions = ['categorical_crossentropy', 'mean_squared_error']
49 batch_sizes = [32, 64, 128]
50 learning_rates = [0.001, 0.01, 0.1]
51
52 results = []
53
54 for activation_func in activation_functions:
55     for loss_func in loss_functions:
56         for batch_size in batch_sizes:
57             for learning_rate in learning_rates:
58                 history, model = create_and_train_model(activation_func, loss_func, batch_size, learni
59                 accuracy = history.history['accuracy'][-1]
60                 val_accuracy = history.history['val_accuracy'][-1]
61                 results.append((activation_func, loss_func, batch_size, learning_rate, accuracy, val_a
62
63 # Plot the accuracy for train and test data for each combination
64 plt.figure(figsize=(12, 8))
65 for result in results:
66     activation_func, loss_func, batch_size, learning_rate, accuracy, val_accuracy, history, model = re
67     label = f'{activation_func}_{loss_func}_{batch_size}_{learning_rate}'
68     plt.plot(history.history['accuracy'], label=f'{label}_train', linestyle='-')
69     plt.plot(history.history['val_accuracy'], label=f'{label}_test', linestyle='--')
70
71 plt.xlabel('Epochs')
72 plt.ylabel('Accuracy')
73 plt.legend(loc='lower right')
74 plt.show()
75
76 # Evaluate the model's performance using different performance metrics
77 for result in results:
78     activation_func, loss_func, batch_size, learning_rate, accuracy, val_accuracy, history, model = re
```

```
79     y_pred = model.predict(test_images)
80     y_pred_classes = np.argmax(y_pred, axis=1)
81     y_true = np.argmax(test_labels_encoded, axis=1)
82
83     print(f'Model with {activation_func} activation, {loss_func} loss, batch size {batch_size}, learni
84
85     # Classification Report
86     print(classification_report(y_true, y_pred_classes))
87
88     # Confusion Matrix
89     cm = confusion_matrix(y_true, y_pred_classes)
90     print("Confusion Matrix:")
91     print(cm)
```





313/313 [=====] - 1s 3ms/step

Model with relu activation, categorical\_crossentropy loss, batch size 32, learning r:

	precision	recall	f1-score	support
0	0.96	0.98	0.97	980
1	0.98	0.98	0.98	1135
2	0.95	0.94	0.95	1032
3	0.94	0.94	0.94	1010
4	0.93	0.95	0.94	982
5	0.94	0.94	0.94	892
6	0.95	0.96	0.96	958
7	0.96	0.95	0.95	1028
8	0.94	0.93	0.93	974
9	0.94	0.92	0.93	1009
accuracy			0.95	10000
macro avg	0.95	0.95	0.95	10000
weighted avg	0.95	0.95	0.95	10000

Confusion Matrix:

```
[[ 963  0  1  1  0  4  6  2  3  0]
 [  0 1114  2  3  0  1  5  1  9  0]
 [  6  0 973  7 11  3  4 10 17  1]
 [  0  2 10 953  1 14  4  9 10  7]
 [  2  3  4  1 931  0  6  2  4 29]
 [  6  1  0 19  4 838 11  1  9  3]
 [ 10  3  4  0  5 10 921  1  4  0]
 [  3  5 18  6  5  1  0 973  2 15]
 [  7  2  8 11  8 11  8  8 907  4]
 [  6  6  2  9 32  8  1 10  3 932]]
```

313/313 [=====] - 1s 3ms/step

Model with relu activation, categorical\_crossentropy loss, batch size 32, learning rate 0.001

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

Confusion Matrix:

```
[[ 969  1  1  0  1  2  4  1  1  0]
 [  0 1122  3  1  0  0  3  1  5  0]
 [  6  2 997  4  5  0  3  9  6  0]
 [  0  0  5 981  0 10  0  5  4  5]
 [  1  0  2  1 961  1  3  1  0 12]
 [  5  0  0 11  0 864  6  1  2  3]
 [  5  3  1  1  8  3 934  1  2  0]
 [  1  4 10  3  3  0  0 994  2 11]
 [  2  0  2  5  1  6  5  4 943  6]
 [  5  3  0  3 10  0  0  4  2 982]]
```

313/313 [=====] - 1s 3ms/step

Model with relu activation, categorical\_crossentropy loss, batch size 32, learning rate 0.001

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

Confusion Matrix:

```
[[ 974  1  1  0  0  1  1  1  1  0]
 [  0 1126  3  1  0  1  2  0  2  0]
 [  4  1 1010  4  1  1  0  9  2  0]
 [  0  0  0 998  0  1  0  4  5  2]
 [  2  1  2  1 954  0  3  2  2 15]
 [  2  1  0 10  0 870  4  1  2  2]
 [  4  1  1  1  6  6 938  0  1  0]
 [  1  1  7  1  1  0  0 1011  2  4]
 [  3  1  2  6  1  2  1  3 952  3]
 [  2  2  0  5 13  3  0  4  2 978]]
```

313/313 [=====] - 1s 4ms/step

Model with relu activation, categorical\_crossentropy loss, batch size 64, learning rate 0.001

	precision	recall	f1-score	support
0	0.95	0.98	0.96	980
1	0.97	0.98	0.97	1135
2	0.95	0.92	0.93	1032
3	0.92	0.92	0.92	1010
4	0.91	0.93	0.92	982
5	0.92	0.88	0.90	892
6	0.93	0.96	0.94	958
7	0.93	0.93	0.93	1028
8	0.91	0.90	0.91	974
9	0.92	0.91	0.91	1009
accuracy			0.93	10000
macro avg	0.93	0.93	0.93	10000
weighted avg	0.93	0.93	0.93	10000

Confusion Matrix:

```
[[ 959  1  0  0  0  6 11  1  2  0]
 [  0 1107  2  3  1  1  5  2 14  0]
 [  9  2 946 17 13  4  8 20 13  0]
 [  4  0 12 931  1 24  4 11 13 10]
 [  1  3  3  1 918  0 14  0  7 35]
 [  8  4  6 27 10 787 13  6 24  7]
 [ 11  3  1  1 10 11 915  2  4  0]
 [  2 17 20  4  9  0  0 953  2 21]
 [  6  4  6 17 10 19 12 11 877 12]
 [ 12  6  0 13 33  7  0 16  4 918]]
```

313/313 [=====] - 1s 3ms/step  
Model with relu activation, categorical\_crossentropy loss, batch size 64, learning rate 0.001

	precision	recall	f1-score	support
0	0.98	0.98	0.98	980
1	0.99	0.99	0.99	1135
2	0.96	0.97	0.97	1032
3	0.96	0.97	0.97	1010
4	0.96	0.97	0.97	982
5	0.97	0.97	0.97	892
6	0.97	0.97	0.97	958
7	0.97	0.96	0.96	1028
8	0.95	0.97	0.96	974
9	0.97	0.95	0.96	1009
accuracy			0.97	10000
macro avg	0.97	0.97	0.97	10000
weighted avg	0.97	0.97	0.97	10000

Confusion Matrix:

```
[[ 964  0  1  0  0  2  8  1  3  1]
 [  0 1120  2  2  0  0  5  1  5  0]
 [  5  1 1000  5  5  0  2  5  9  0]
 [  0  0  6 978  0  9  0  4  8  5]
 [  1  0  9  0 957  0  2  2  2  9]
 [  3  0  0  6  2 863  8  2  6  2]
 [  5  4  1  0  6  8 925  2  7  0]
 [  1  9 14  7  4  0  0 982  2  9]
 [  2  0  4  8  5  5  4  2 940  4]
 [  3  3  0  9 21  2  1  7  7 956]]
```

313/313 [=====] - 1s 3ms/step  
Model with relu activation, categorical\_crossentropy loss, batch size 64, learning rate 0.001

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

Confusion Matrix:

```
[[ 967  0  1  0  0  1  3  2  4  2]
 [  1 1125  2  0  0  1  2  1  3  0]
 [  4  1 1011  1  1  0  2  4  8  0]
 [  0  0  4 993  0  3  0  3  1  6]
 [  1  0  0  1 957  0  5  0  1 17]
 [  2  0  0  6  1 875  3  1  3  1]
 [  3  3  0  1  6  7 936  0  2  0]
 [  0  1  8  2  2  0  0 1007  3  5]
 [  0  0  2  6  2  5  2  3 951  3]
 [  4  2  1  5  6  4  1  5  2 979]]
```

313/313 [=====] - 1s 3ms/step  
Model with relu activation, categorical\_crossentropy loss, batch size 128, learning rate 0.001

	precision	recall	f1-score	support
--	-----------	--------	----------	---------

0	0.94	0.97	0.96	980
1	0.95	0.98	0.96	1135
2	0.91	0.89	0.90	1032
3	0.91	0.92	0.91	1010
4	0.89	0.92	0.91	982
5	0.90	0.87	0.89	892
6	0.93	0.94	0.93	958
7	0.93	0.90	0.91	1028
8	0.91	0.88	0.90	974
9	0.90	0.88	0.89	1009
accuracy			0.92	10000
macro avg	0.92	0.92	0.92	10000
weighted avg	0.92	0.92	0.92	10000

Confusion Matrix:

```
[[ 955  1  1  2  0  9  7  2  3  0]
 [  0 1113  2  3  1  0  4  0 12  0]
 [ 10  7 918 14 23  5 13 18 22  2]
 [  3  3 17 926  3 24  4 12 13  5]
 [  1  6  5  2 907  1 13  1  6 40]
 [ 14  4  7 30  9 777 17  9 17  8]
 [ 14  5  6  1  6 14 905  2  4  1]
 [  2 18 34  3  6  3  3 929  1 29]
 [  6  8 10 22 15 25  9 11 859  9]
 [ 10  8  5 18 45  4  3 19  6 891]]
```

313/313 [=====] - 1s 3ms/step

Model with relu activation, categorical\_crossentropy loss, batch size 128, learning r

	precision	recall	f1-score	support
0	0.97	0.98	0.98	980
1	0.99	0.99	0.99	1135
2	0.95	0.96	0.95	1032
3	0.95	0.96	0.95	1010
4	0.96	0.97	0.96	982
5	0.95	0.96	0.95	892
6	0.96	0.96	0.96	958
7	0.97	0.96	0.96	1028
8	0.95	0.95	0.95	974
9	0.96	0.95	0.96	1009
accuracy			0.96	10000
macro avg	0.96	0.96	0.96	10000
weighted avg	0.96	0.96	0.96	10000

Confusion Matrix:

```
[[ 958  0  2  1  0  3  9  2  4  1]
 [  0 1119  5  2  0  1  2  0  6  0]
 [  4  1 986  7  5  1 10  9  9  0]
 [  0  0  8 965  0 18  0  7  9  3]
 [  3  0  7  1 948  0  4  2  3 14]
 [  5  1  1 15  1 853  7  1  5  3]
 [  7  3  0  1  7 11 922  1  6  0]
 [  1  7 19  4  1  0  0 983  3 10]
 [  4  0  4 13  6  5  6  3 926  7]
 [  3  4  1  5 15  3  1  6  8 963]]
```

313/313 [=====] - 1s 3ms/step

Model with relu activation, categorical\_crossentropy loss, batch size 128, learning r

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

Confusion Matrix:

```
[[ 971  0  1  0  0  2  4  1  1  0]
 [  0 1121  2  1  0  2  3  1  5  0]
 [  4  0 1006  6  3  0  3  3  7  0]
 [  2  0  2 990  0  5  0  3  3  5]
 [  4  0  1  1 963  0  3  2  0  8]
 [  3  0  0 11  1 865  4  2  3  3]
 [  4  2  0  0  2  9 936  0  5  0]
 [  2  2  8  3  1  0  0 1003  4  5]
 [  5  0  2  8  3  3  4  2 943  4]
 [  5  2  1  6 12  1  0  4  4 974]]
```

313/313 [=====] - 1s 3ms/step

Model with relu activation, mean\_squared\_error loss, batch size 32, learning rate 0.0

	precision	recall	f1-score	support
--	-----------	--------	----------	---------

0	0.86	0.92	0.89	980
1	0.84	0.98	0.90	1135
2	0.82	0.72	0.76	1032
3	0.78	0.82	0.80	1010
4	0.74	0.80	0.77	982
5	0.74	0.52	0.61	892
6	0.80	0.87	0.83	958
7	0.82	0.81	0.81	1028
8	0.76	0.67	0.71	974
9	0.74	0.79	0.77	1009
accuracy			0.79	10000
macro avg	0.79	0.79	0.79	10000
weighted avg	0.79	0.79	0.79	10000

Confusion Matrix:

```
[[ 901  5 16 11  0 16 21  5  4  1]
 [  0 1112  4  4  1  1  3  2  8  0]
 [ 18 47 740 36 34 10 42 38 52 15]
 [  8 11 27 826  9 42  8 29 27 23]
 [  8  7  4  4 785 10 51  4 11 98]
 [ 37 49 22 97 64 466 34 26 64 33]
 [ 31 14 25  6 25 17 829  2  9  0]
 [ 10 33 16  3 33  6  7 828 20 72]
 [ 16 36 46 53 35 38 33 30 653 34]
 [ 15 11  5 14 72 22  9 51 13 797]]
```

313/313 [=====] - 1s 3ms/step

Model with relu activation, mean\_squared\_error loss, batch size 32, learning rate 0.0

	precision	recall	f1-score	support
0	0.94	0.98	0.96	980
1	0.97	0.98	0.97	1135
2	0.93	0.91	0.92	1032
3	0.93	0.92	0.92	1010
4	0.92	0.94	0.93	982
5	0.93	0.90	0.92	892
6	0.93	0.95	0.94	958
7	0.94	0.92	0.93	1028
8	0.91	0.91	0.91	974
9	0.92	0.91	0.91	1009
accuracy			0.93	10000
macro avg	0.93	0.93	0.93	10000
weighted avg	0.93	0.93	0.93	10000

Confusion Matrix:

```
[[ 956  1  0  0  2  5  8  2  6  0]
 [  0 1111  3  3  0  0  6  1 11  0]
 [  8  2 938 14 10  4 15 14 16 11]
 [  3  1 20 929  4 22  2  9 13  7]
 [  2  2  6  0 920  0 11  2  8 31]
 [  8  0  3 26  3 806 14  5 23  4]
 [ 15  4  5  0  9  9 912  1  3  0]
 [  2 13 22  6 12  3  1 948  1 20]
 [ 13  4  7 14  8 13 15  8 883  9]
 [ 11  9  1 12 28  7  1 22  4 914]]
```

313/313 [=====] - 1s 3ms/step

Model with relu activation, mean\_squared\_error loss, batch size 32, learning rate 0.1

	precision	recall	f1-score	support
0	0.97	0.99	0.98	980
1	0.98	0.99	0.98	1135
2	0.96	0.97	0.96	1032
3	0.97	0.96	0.97	1010
4	0.96	0.97	0.97	982
5	0.97	0.96	0.96	892
6	0.97	0.98	0.97	958
7	0.97	0.96	0.97	1028
8	0.97	0.96	0.96	974
9	0.98	0.95	0.96	1009
accuracy			0.97	10000
macro avg	0.97	0.97	0.97	10000
weighted avg	0.97	0.97	0.97	10000

Confusion Matrix:

```
[[ 967  0  2  2  1  1  3  2  1  1]
 [  0 1120  4  2  1  1  4  0  3  0]
 [  6  1 1005  5  2  0  2  6  5  0]
 [  0  0 11 971  1 12  2  6  4  3]
 [  1  0  6  1 957  0  3  2  4  8]
 [  4  0  2  7  2 856  8  2  8  3]
 [  6  2  2  0  6  4 936  1  1  0]
 [  1 10 14  1  3  1  1 991  2  4]
 [  6  2  5  4  5  5  4  4 935  4]
 [  4  6  1  9 19  4  2  9  1 954]]
```

313/313 [=====] - 1s 4ms/step

Model with relu activation, mean\_squared\_error loss, batch size 64, learning rate 0.1

	precision	recall	f1-score	support
--	-----------	--------	----------	---------

0	0.71	0.88	0.79	980
1	0.76	0.94	0.84	1135
2	0.73	0.65	0.69	1032
3	0.65	0.64	0.65	1010
4	0.62	0.64	0.63	982
5	0.52	0.33	0.41	892
6	0.80	0.78	0.79	958
7	0.71	0.81	0.76	1028
8	0.57	0.49	0.53	974
9	0.62	0.57	0.60	1009
accuracy			0.68	10000
macro avg	0.67	0.68	0.67	10000
weighted avg	0.67	0.68	0.67	10000

Confusion Matrix:

```
[[ 863  7 16  8  9 14 29 17 15  2]
 [  0 1067 25  8 12  1  4  2 10  6]
 [ 41  60 671 63 21 20 42 55 44 15]
 [ 26 33 46 650 11 102 20 40 73  9]
 [ 20 23 25 20 633 16 27 10 29 179]
 [115 91 22 81 64 298 28 35 145 13]
 [ 47 20 42  3 52 15 750  1  3 25]
 [ 14 34 15 13 29 18  5 834 14 52]
 [ 44 38 46 137 67 66 17 32 478 49]
 [ 44 27  9 19 123 19 11 142 35 580]]
```

313/313 [=====] - 1s 3ms/step

Model with relu activation, mean\_squared\_error loss, batch size 64, learning rate 0.001

	precision	recall	f1-score	support
0	0.92	0.97	0.94	980
1	0.96	0.98	0.97	1135
2	0.93	0.88	0.90	1032
3	0.92	0.90	0.91	1010
4	0.89	0.94	0.92	982
5	0.90	0.87	0.89	892
6	0.92	0.95	0.93	958
7	0.92	0.91	0.92	1028
8	0.88	0.87	0.87	974
9	0.90	0.88	0.89	1009

accuracy			0.92	10000
macro avg	0.91	0.92	0.91	10000
weighted avg	0.92	0.92	0.92	10000

Confusion Matrix:

```
[[ 950  0  0  3  0  5 12  1  7  2]
 [  0 1111 2  4  1  1  6  0 10  0]
 [ 18  4 910 11 21  1 16 17 30  4]
 [  2  1 24 904  1 26  4 18 22  8]
 [  1  2  3  2 927  0 10  1  5 31]
 [ 16  1  6 26 11 780 17  7 17 11]
 [ 19  3  6  0  9  8 908  0  5  0]
 [  6 20 19  1 13  3  1 935  6 24]
 [  7  4 10 28  9 28 16 13 843 16]
 [ 12  6  1  6 44 13  2 21 12 892]]
```

313/313 [=====] - 1s 3ms/step

Model with relu activation, mean\_squared\_error loss, batch size 64, learning rate 0.001

	precision	recall	f1-score	support
0	0.97	0.99	0.98	980
1	0.99	0.99	0.99	1135
2	0.95	0.95	0.95	1032
3	0.95	0.96	0.96	1010
4	0.96	0.96	0.96	982
5	0.97	0.95	0.96	892
6	0.97	0.97	0.97	958
7	0.96	0.96	0.96	1028
8	0.94	0.95	0.95	974
9	0.96	0.94	0.95	1009

accuracy			0.96	10000
macro avg	0.96	0.96	0.96	10000
weighted avg	0.96	0.96	0.96	10000

Confusion Matrix:

```
[[ 967  0  1  2  0  1  4  1  3  1]
 [  0 1120 5  0  0  3  2  1  4  0]
 [  6  0 984  5  5  2  5  9 13  3]
 [  0  1 11 971  0  6  0 12  7  2]
 [  2  0  5  0 944  1  9  1  3 17]
 [  3  1  2 13  2 849  7  3  7  5]
 [  8  2  2  0  4  6 931  0  5  0]
 [  1  5 18  4  5  1  0 982  1 11]
 [  4  0  8 14  5  5  5  6 923  4]
 [  7  4  0  9 19  3  0  5 13 949]]
```

313/313 [=====] - 1s 3ms/step

Model with relu activation, mean\_squared\_error loss, batch size 128, learning rate 0.001

0	0.59	0.76	0.67	980
1	0.70	0.84	0.76	1135
2	0.32	0.22	0.26	1032
3	0.46	0.41	0.43	1010
4	0.41	0.45	0.43	982
5	0.21	0.11	0.14	892
6	0.60	0.71	0.65	958
7	0.61	0.70	0.65	1028
8	0.50	0.52	0.51	974
9	0.31	0.29	0.30	1009
accuracy			0.51	10000
macro avg	0.47	0.50	0.48	10000
weighted avg	0.48	0.51	0.49	10000

Confusion Matrix:

[	747	9	19	38	32	13	63	11	18	30]
[	3	949	81	1	34	23	5	4	34	1]
[	127	87	226	29	58	76	134	79	100	116]
[	65	48	97	413	33	70	16	115	95	58]
[	32	48	32	51	442	40	73	16	27	221]
[	152	32	70	154	107	97	52	18	137	73]
[	55	24	4	34	27	54	682	15	13	50]
[	14	62	58	17	39	38	3	715	46	36]
[	39	65	75	62	79	25	37	40	504	48]
[	25	40	38	97	219	27	79	154	39	291]]

313/313 [=====] - 1s 3ms/step

Model with relu activation, mean\_squared\_error loss, batch size 128, learning rate 0

	precision	recall	f1-score	support
0	0.89	0.96	0.92	980
1	0.91	0.98	0.94	1135
2	0.90	0.84	0.87	1032
3	0.88	0.88	0.88	1010
4	0.88	0.87	0.87	982
5	0.86	0.79	0.82	892
6	0.89	0.92	0.91	958
7	0.90	0.88	0.89	1028
8	0.87	0.82	0.84	974
9	0.85	0.87	0.86	1009
accuracy			0.88	10000
macro avg	0.88	0.88	0.88	10000
weighted avg	0.88	0.88	0.88	10000

Confusion Matrix:

[	940	0	5	5	0	9	12	2	6	1]
[	0	1110	4	5	0	2	5	0	9	0]
[	25	16	869	24	13	3	18	21	36	7]
[	9	9	19	886	2	30	4	11	18	22]
[	7	11	9	1	853	4	14	7	9	67]
[	21	15	3	43	24	705	29	22	25	5]
[	24	8	7	1	15	13	883	2	4	1]
[	6	22	37	6	12	2	1	908	4	30]
[	14	17	11	29	10	44	17	11	796	25]
[	14	12	2	10	42	11	5	26	12	875]]

313/313 [=====] - 1s 3ms/step

Model with relu activation, mean\_squared\_error loss, batch size 128, learning rate 0

	precision	recall	f1-score	support
0	0.96	0.98	0.97	980
1	0.98	0.99	0.98	1135
2	0.95	0.94	0.94	1032
3	0.95	0.94	0.94	1010
4	0.94	0.96	0.95	982
5	0.95	0.94	0.94	892
6	0.95	0.96	0.96	958
7	0.96	0.95	0.95	1028
8	0.95	0.94	0.95	974
9	0.95	0.94	0.94	1009
accuracy			0.95	10000
macro avg	0.95	0.95	0.95	10000
weighted avg	0.95	0.95	0.95	10000

Confusion Matrix:

[	962	0	0	1	1	4	7	3	2	0]
[	0	1120	2	2	0	0	5	0	6	0]
[	10	1	972	11	9	1	7	9	10	2]
[	2	2	18	946	2	16	1	10	11	2]
[	1	1	2	0	944	0	11	0	3	20]
[	7	2	0	16	2	836	11	5	6	7]
[	4	3	6	1	9	10	924	0	1	0]
[	1	8	21	2	7	0	0	974	0	15]
[	7	1	5	7	12	6	9	8	913	6]
[	7	5	1	10	21	5	2	8	6	944]]

313/313 [=====] - 1s 3ms/step

Model with tanh activation, categorical\_crossentropy loss, batch size 32, learning r:

	precision	recall	f1-score	support
--	-----------	--------	----------	---------

0	0.95	0.98	0.96	980
1	0.96	0.98	0.97	1135
2	0.94	0.91	0.92	1032
3	0.93	0.91	0.92	1010
4	0.91	0.94	0.92	982
5	0.91	0.88	0.90	892
6	0.93	0.95	0.94	958
7	0.93	0.93	0.93	1028
8	0.90	0.88	0.89	974
9	0.91	0.90	0.90	1009

accuracy			0.93	10000
macro avg	0.93	0.93	0.93	10000
weighted avg	0.93	0.93	0.93	10000

Confusion Matrix:

```
[[ 956  0  0  1  0  8 11  2  2  0]
 [  0 1112  2  2  0  2  4  1 12  0]
 [  8  4 935 11 10  4 13 13 27  7]
 [  3  3 16 924  2 20  5 12 15 10]
 [  2  4  5  0 920  0 10  2  6 33]
 [ 10  2  3 23 10 788 16  8 24  8]
 [  8  3  4  1 11 13 914  1  3  0]
 [  3 13 21  4 10  0  0 951  1 25]
 [  9  9  7 19 10 23 14 11 860 12]
 [ 12  7  1 13 41  5  1 19  5 905]]
```

313/313 [=====] - 1s 3ms/step  
Model with tanh activation, categorical\_crossentropy loss, batch size 32, learning rate 0.001

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

accuracy			0.97	10000
macro avg	0.97	0.97	0.97	10000
weighted avg	0.97	0.97	0.97	10000

Confusion Matrix:

```
[[ 970  0  0  1  0  3  3  1  2  0]
 [  0 1125  3  0  0  1  2  1  3  0]
 [  6  2 999  2  0  2  2  8 11  0]
 [  0  0  4 986  0  6  0  7  4  3]
 [  1  0  4  1 946  0  6  2  2 20]
 [  4  1  1 12  2 857  7  0  6  2]
 [  7  3  0  2  2  4 935  2  3  0]
 [  1  7 11  4  1  0  0 992  1 11]
 [  3  1  5  7  4  4  4  4 937  5]
 [  4  4  1  8 10  2  0  3  1 976]]
```

313/313 [=====] - 1s 4ms/step  
Model with tanh activation, categorical\_crossentropy loss, batch size 32, learning rate 0.001

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

accuracy			0.98	10000
macro avg	0.98	0.98	0.98	10000
weighted avg	0.98	0.98	0.98	10000

Confusion Matrix:

```
[[ 966  0  1  2  1  0  5  2  2  1]
 [  1 1125  3  0  0  1  2  1  2  0]
 [  2  1 1016  3  2  0  0  4  4  0]
 [  0  1  5 989  0  4  0  5  3  3]
 [  2  1  4  1 947  0  3  3  1 20]
 [  2  0  0 12  1 863  3  1  5  5]
 [  6  3  1  1  6  3 932  0  5  1]
 [  0  3 12  1  0  0  0 1010  2  0]
 [  4  0  3  3  2  0  0  3 957  2]
 [  1  3  0  4  2  1  2 11  1 984]]
```

313/313 [=====] - 1s 3ms/step  
Model with tanh activation, categorical\_crossentropy loss, batch size 64, learning rate 0.001

	precision	recall	f1-score	support
--	-----------	--------	----------	---------



0	0.95	0.97	0.96	980
1	0.95	0.97	0.96	1135
2	0.92	0.89	0.91	1032
3	0.91	0.90	0.90	1010
4	0.90	0.93	0.92	982
5	0.89	0.86	0.87	892
6	0.92	0.94	0.93	958
7	0.92	0.91	0.91	1028
8	0.89	0.87	0.88	974
9	0.90	0.89	0.89	1009
accuracy			0.91	10000
macro avg	0.91	0.91	0.91	10000
weighted avg	0.91	0.91	0.91	10000

Confusion Matrix:

```
[[ 948  0  1  2  0 11 12  3  3  0]
 [  0 1106  2  6  0  2  4  0 15  0]
 [ 14 10 916 13 12  3 16 15 28  5]
 [  6  2  20 913  1 25  3 16 19  5]
 [  1  3  6  1 913  1 13  2  4 38]
 [  9  5  5 35  9 767 20  9 25  8]
 [ 12  4  8  0 12 14 905  1  2  0]
 [  2 15 27  4 10  0  1 932  0 37]
 [ 10 11  6 22 13 33 12 10 845 12]
 [ 14  6  1 12 40  8  1 27  4 896]]
```

313/313 [=====] - 1s 3ms/step  
Model with tanh activation, categorical\_crossentropy loss, batch size 64, learning rate 0.001

	precision	recall	f1-score	support
0	0.96	0.99	0.97	980
1	0.98	0.99	0.99	1135
2	0.96	0.95	0.96	1032
3	0.94	0.96	0.95	1010
4	0.97	0.96	0.97	982
5	0.97	0.94	0.96	892
6	0.97	0.97	0.97	958
7	0.96	0.95	0.96	1028
8	0.96	0.96	0.96	974
9	0.96	0.95	0.95	1009
accuracy			0.96	10000
macro avg	0.96	0.96	0.96	10000
weighted avg	0.96	0.96	0.96	10000

Confusion Matrix:

```
[[ 966  0  2  1  0  3  5  1  2  0]
 [  0 1123  2  1  0  1  4  1  3  0]
 [  6  3 985 10  4  0  7  7  8  2]
 [  0  1 12 968  1  7  1 11  3  6]
 [  1  0  5  1 946  0  5  2  5 17]
 [  7  0  1 21  3 842  7  4  5  2]
 [ 10  2  0  0  4  9 927  1  5  0]
 [  2  5 17  4  2  2  0 980  2 14]
 [  4  1  5  9  2  4  2  5 939  3]
 [  6  6  1 10 11  2  1  5  6 961]]
```

313/313 [=====] - 1s 3ms/step  
Model with tanh activation, categorical\_crossentropy loss, batch size 64, learning rate 0.001

	precision	recall	f1-score	support
0	0.98	0.99	0.99	980
1	0.99	0.99	0.99	1135
2	0.97	0.98	0.97	1032
3	0.95	0.98	0.97	1010
4	0.98	0.97	0.98	982
5	0.98	0.97	0.97	892
6	0.98	0.98	0.98	958
7	0.97	0.98	0.98	1028
8	0.99	0.95	0.97	974
9	0.97	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

Confusion Matrix:

```
[[ 973  1  0  0  0  0  1  1  3  1]
 [  0 1127  4  0  0  1  0  1  2  0]
 [  5  2 1007  5  1  1  3  4  3  1]
 [  0  0  6 994  0  4  0  3  0  3]
 [  1  0  0  1 953  0  6  3  2 16]
 [  3  0  0 16  1 863  4  1  1  3]
 [  4  3  3  1  4  5 937  0  1  0]
 [  0  4  7  5  0  0  0 1009  1  2]
 [  3  0 10 17  2  3  3  6 928  2]
 [  4  4  1 10  9  2  0 13  1 965]]
```

313/313 [=====] - 1s 3ms/step  
Model with tanh activation, categorical\_crossentropy loss, batch size 128, learning rate 0.001

	precision	recall	f1-score	support
0	0.93	0.97	0.95	980

1	0.94	0.97	0.96	1135
2	0.91	0.86	0.89	1032
3	0.89	0.89	0.89	1010
4	0.88	0.91	0.90	982
5	0.88	0.82	0.85	892
6	0.91	0.94	0.92	958
7	0.90	0.91	0.91	1028
8	0.87	0.83	0.85	974
9	0.88	0.87	0.87	1009
accuracy			0.90	10000
macro avg	0.90	0.90	0.90	10000
weighted avg	0.90	0.90	0.90	10000

Confusion Matrix:

```
[[ 949  0  3  2  1  7 12  2  4  0]
 [  0 1104  2  3  1  3  3  1 18  0]
 [ 11 12 890 19 18  2 14 26 34  6]
 [  6  3  23 897  1 29  5 20 13 13]
 [  0  3  6  2 896  1 19  2  6 47]
 [ 11  8  7 47 14 734 19 10 33  9]
 [ 18  5  7  2 11 12 900  1  2  0]
 [  3 14 26  3  9  2  1 935  2 33]
 [  8 21 10 24 14 35 18 16 813 15]
 [ 13  7  3 12 52 10  3 22  8 879]]
```

313/313 [=====] - 1s 3ms/step

Model with tanh activation, categorical\_crossentropy loss, batch size 128, learning r

	precision	recall	f1-score	support
0	0.95	0.99	0.97	980
1	0.98	0.99	0.98	1135
2	0.96	0.94	0.95	1032
3	0.94	0.95	0.94	1010
4	0.94	0.95	0.95	982
5	0.94	0.92	0.93	892
6	0.95	0.96	0.96	958
7	0.95	0.95	0.95	1028
8	0.93	0.93	0.93	974
9	0.95	0.92	0.94	1009
accuracy			0.95	10000
macro avg	0.95	0.95	0.95	10000
weighted avg	0.95	0.95	0.95	10000

Confusion Matrix:

```
[[ 966  0  0  2  0  4  6  1  1  0]
 [  0 1121  2  2  0  2  3  2  3  0]
 [  9  2 972  8  7  1  8  7 16  2]
 [  1  1  9 955  1 12  2 13 13  3]
 [  1  1  3  2 937  0  8  1  5 24]
 [  8  3  0 22  2 822 10  5 17  3]
 [  9  3  3  0  6 12 921  1  3  0]
 [  3  8 19  5  5  1  0 972  2 13]
 [  6  4  4  9  9 15 11  7 905  4]
 [ 10  6  1 10 25  5  1 12  6 933]]
```

313/313 [=====] - 1s 3ms/step

Model with tanh activation, categorical\_crossentropy loss, batch size 128, learning r

	precision	recall	f1-score	support
0	0.98	0.98	0.98	980
1	0.99	0.99	0.99	1135
2	0.99	0.97	0.98	1032
3	0.93	0.99	0.96	1010
4	0.98	0.97	0.97	982
5	0.97	0.97	0.97	892
6	0.98	0.97	0.98	958
7	0.97	0.97	0.97	1028
8	0.98	0.94	0.96	974
9	0.96	0.96	0.96	1009
accuracy			0.97	10000
macro avg	0.97	0.97	0.97	10000
weighted avg	0.97	0.97	0.97	10000

Confusion Matrix:

```
[[ 965  0  0  3  0  2  5  2  2  1]
 [  0 1124  2  4  0  1  1  1  2  0]
 [  2  0 999 15  2  0  3  5  5  1]
 [  0  0  1 996  0  4  0  6  2  1]
 [  0  1  2  1 948  0  6  6  0 18]
 [  4  0  0 12  0 865  3  0  3  5]
 [  6  3  0  2  5  7 932  1  2  0]
 [  1  3  6  6  1  0  0 1002  1  8]
 [  3  1  1 31  1  8  2  5 919  3]
 [  2  3  1  5 10  7  0  7  2 972]]
```

313/313 [=====] - 1s 3ms/step

Model with tanh activation, mean\_squared\_error loss, batch size 32, learning rate 0.6

	precision	recall	f1-score	support
0	0.85	0.95	0.89	980

	1	0.81	0.96	0.88	1135
	2	0.82	0.73	0.77	1032
	3	0.76	0.80	0.78	1010
	4	0.78	0.84	0.81	982
	5	0.79	0.58	0.67	892
	6	0.81	0.89	0.85	958
	7	0.81	0.87	0.84	1028
	8	0.81	0.63	0.71	974
	9	0.79	0.73	0.76	1009
accuracy				0.80	10000
macro avg	0.80	0.80	0.80		10000
weighted avg	0.80	0.80	0.80		10000

Confusion Matrix:

```
[[ 927  2  5  8  2  6 24  5  1  0]
 [  0 1085 17  4  3  9  4  2 11  0]
 [ 19  52 758 50 17  3 37 37 53  6]
 [ 10 12  32 812  1 52 11 42 24 14]
 [  2 11 11  9 829  4 26  6  6 78]
 [ 69  54  28  80  29 520 49 23 21 19]
 [ 19 14 12  2  32 18 856  5  0  0]
 [  6 35 21  4 12  0  4 893  8 45]
 [ 24  54 31 87 30 39 43 18 613 35]
 [ 19 18 13 17 106 11  1  67 18 739]]
```

313/313 [=====] - 1s 4ms/step  
Model with tanh activation, mean\_squared\_error loss, batch size 32, learning rate 0.6

		precision	recall	f1-score	support
	0	0.93	0.98	0.95	980
	1	0.96	0.98	0.97	1135
	2	0.92	0.88	0.90	1032
	3	0.91	0.91	0.91	1010
	4	0.89	0.94	0.91	982
	5	0.91	0.83	0.87	892
	6	0.92	0.95	0.93	958
	7	0.90	0.91	0.90	1028
	8	0.90	0.86	0.88	974
	9	0.89	0.88	0.88	1009
accuracy				0.91	10000
macro avg	0.91	0.91	0.91		10000
weighted avg	0.91	0.91	0.91		10000

Confusion Matrix:

```
[[ 961  0  2  1  0  3  6  3  4  0]
 [  0 1107  2  4  1  2  4  1 14  0]
 [ 16  3  909 11 16  3 15 20 30  9]
 [  3  2  26 917  2 18  3 15 12 12]
 [  1  1  4  0 920  0 14  2  3 37]
 [ 15  2  5 35 20 744 20 16 27  8]
 [ 17  3  3  1 10  9 910  3  2  0]
 [  3 18 19  3 12  0  1 936  1 35]
 [  7 11 11 23 14 26 17 16 838 11]
 [ 14  7  2 11 42 10  0 30  5 888]]
```

313/313 [=====] - 1s 4ms/step  
Model with tanh activation, mean\_squared\_error loss, batch size 32, learning rate 0.1

		precision	recall	f1-score	support
	0	0.96	0.98	0.97	980
	1	0.98	0.99	0.98	1135
	2	0.95	0.95	0.95	1032
	3	0.96	0.95	0.95	1010
	4	0.97	0.96	0.96	982
	5	0.96	0.94	0.95	892
	6	0.96	0.97	0.96	958
	7	0.94	0.95	0.95	1028
	8	0.95	0.95	0.95	974
	9	0.96	0.94	0.95	1009
accuracy				0.96	10000
macro avg	0.96	0.96	0.96		10000
weighted avg	0.96	0.96	0.96		10000

Confusion Matrix:

```
[[ 964  0  1  1  0  3  6  4  1  0]
 [  0 1120  4  2  0  1  4  1  3  0]
 [  6  4  977  6  7  1  5 11 12  3]
 [  0  0 14 957  2 14  1 11  8  3]
 [  1  0  4  0 941  1 11  3  3 18]
 [  8  1  1  8  3 838  9  5 13  6]
 [  6  2  1  1  4  6 931  2  5  0]
 [  2 11 18  5  1  1  0 977  2 11]
 [  6  0  2 10  5  5  6  9 928  3]
 [  6  5  2 12  9  4  1 11  7 952]]
```

313/313 [=====] - 1s 3ms/step  
Model with tanh activation, mean\_squared\_error loss, batch size 64, learning rate 0.6

		precision	recall	f1-score	support
	0	0.78	0.84	0.81	980
	1	0.77	0.86	0.82	1135

	1	0.77	0.96	0.85	1135
	2	0.78	0.70	0.74	1032
	3	0.65	0.76	0.70	1010
	4	0.70	0.79	0.74	982
	5	0.55	0.10	0.17	892
	6	0.71	0.89	0.79	958
	7	0.74	0.78	0.76	1028
	8	0.66	0.66	0.66	974
	9	0.73	0.62	0.67	1009
accuracy				0.72	10000
macro avg	0.71	0.71	0.69		10000
weighted avg	0.71	0.72	0.70		10000

Confusion Matrix:

```
[[ 825  3  5 12  6 14 84 15 12  4]
 [  0 1088  9 11  2  0  7  2 16  0]
 [ 20  63 727 25 28  7 54 43 51 14]
 [ 21  35  52 767 13 20 23 18 35 26]
 [  5  20  10  5 774  3 65 14 16 70]
 [ 109  78  20 255  78 89 54 46 129 34]
 [ 25 11 16 12 15  3 857  2 16  1]
 [ 13 50 30  9 25  6 10 799 28 58]
 [ 23 51 56 68 29 17 38 23 645 24]
 [ 20 18 12 23 143  4  9 125 25 630]]
```

313/313 [=====] - 1s 3ms/step  
Model with tanh activation, mean\_squared\_error loss, batch size 64, learning rate 0.001

	precision	recall	f1-score	support
0	0.93	0.97	0.95	980
1	0.95	0.97	0.96	1135
2	0.89	0.86	0.87	1032
3	0.90	0.90	0.90	1010
4	0.86	0.92	0.89	982
5	0.88	0.81	0.84	892
6	0.91	0.94	0.92	958
7	0.91	0.91	0.91	1028
8	0.87	0.84	0.85	974
9	0.89	0.87	0.88	1009
accuracy			0.90	10000
macro avg	0.90	0.90	0.90	10000
weighted avg	0.90	0.90	0.90	10000

Confusion Matrix:

```
[[ 952  0  6  1  1  7  9  1  3  0]
 [  0 1098  1  5  2  1  6  2 20  0]
 [ 15  8 887 21 20  2 15 21 36  7]
 [  5  1  20 907  2 32  4 12 16 11]
 [  1  5  5  0 906  1 17  1  5 41]
 [ 15  4 13 42 19 722 18 15 35  9]
 [ 12  4  9  1 10 19 898  2  3  0]
 [  5 14 31  4 12  1  0 933  1 27]
 [ 11 15 17 20 21 24 16 20 817 13]
 [ 13  7  7 12 55 13  2 20  3 877]]
```

313/313 [=====] - 1s 3ms/step  
Model with tanh activation, mean\_squared\_error loss, batch size 64, learning rate 0.1

	precision	recall	f1-score	support
0	0.95	0.98	0.97	980
1	0.97	0.99	0.98	1135
2	0.95	0.93	0.94	1032
3	0.93	0.93	0.93	1010
4	0.93	0.96	0.95	982
5	0.95	0.90	0.92	892
6	0.94	0.96	0.95	958
7	0.94	0.94	0.94	1028
8	0.93	0.92	0.93	974
9	0.94	0.93	0.93	1009
accuracy			0.94	10000
macro avg	0.94	0.94	0.94	10000
weighted avg	0.94	0.94	0.94	10000

Confusion Matrix:

```
[[ 965  0  1  1  0  3  7  1  2  0]
 [  0 1120  3  3  0  1  4  2  2  0]
 [  8  7 957  8  9  1 11 11 18  2]
 [  1  0 16 941  2 16  1 12 14  7]
 [  1  1  2  2 940  0 12  2  2 20]
 [  8  2  2 22  7 803 16  5 18  9]
 [  8  3  4  3  8  8 921  1  2  0]
 [  2  7 21  2  7  2  0 970  0 17]
 [  7  5  4 15 11 10  7 13 897  5]
 [ 12  5  0 11 23  4  2 10  7 935]]
```

313/313 [=====] - 1s 3ms/step  
Model with tanh activation, mean\_squared\_error loss, batch size 128, learning rate 0.1

		precision	recall	f1-score	support
	0	0.68	0.63	0.65	980
	1	0.64	0.96	0.77	1135

```

-      -      -      -      -
2      0.68      0.69      0.69      1032
3      0.56      0.58      0.57      1010
4      0.57      0.51      0.54      982
5      0.38      0.26      0.31      892
6      0.63      0.54      0.58      958
7      0.63      0.72      0.67      1028
8      0.62      0.52      0.57      974
9      0.48      0.48      0.48      1009

accuracy      0.60      10000
macro avg      0.59      0.59      0.58      10000
weighted avg      0.59      0.60      0.59      10000

```

Confusion Matrix:

```

[[ 616  37  22  55  8  93  39  50  23  37]
 [  0 1084  25  2  12  0  4  1  6  1]
 [ 23  95 714  46  23  27  46  24  33  1]
 [ 34 100  50 589  24  89  23  38  19  44]
 [  3  33  4  15 505  53  57  60  20 232]
 [100  52  24 190  52 235  42  41  86  70]
 [ 35 104 117  12  50  37 519  23  59  2]
 [ 10  82  41  3  26  22  9 741  10  84]
 [ 58  82  44 114  36  25  50 12 504  49]
 [ 30  20  6  27 146  32  29 181  50 488]]

```

313/313 [=====] - 1s 3ms/step

Model with tanh activation, mean\_squared\_error loss, batch size 128, learning rate 0

```

precision      recall      f1-score      support

0      0.89      0.96      0.92      980
1      0.91      0.96      0.94      1135
2      0.89      0.83      0.86      1032
3      0.86      0.87      0.86      1010
4      0.84      0.91      0.87      982
5      0.87      0.77      0.82      892
6      0.87      0.91      0.89      958
7      0.89      0.89      0.89      1028
8      0.85      0.79      0.82      974
9      0.87      0.84      0.86      1009

accuracy      0.88      10000
macro avg      0.87      0.87      0.87      10000
weighted avg      0.88      0.88      0.87      10000

```

Confusion Matrix:

```

[[ 937  1  6  5  1 10 14  1  5  0]
 [  0 1092  4  5  1  3  6  2 22  0]
 [ 22  22 859  21 18  2 21 26 35  6]
 [  7  5  22 876  2 39 14 15 17 13]
 [  3  7  5  2 893  1 12  6  8  45]
 [ 23 12 10  43 29 689 34 16 25 11]
 [ 27  3  4  2 20 16 876  1  9  0]
 [  3 19 28 12 15  0  4 913  1 33]
 [ 19 25 15  44 18 26 22 16 771 18]
 [ 12  8  9 13 69  9  7 25  9 848]]

```

313/313 [=====] - 1s 3ms/step

Model with tanh activation, mean\_squared\_error loss, batch size 128, learning rate 0

```

precision      recall      f1-score      support

0      0.95      0.98      0.96      980
1      0.97      0.98      0.97      1135
2      0.94      0.91      0.92      1032
3      0.92      0.91      0.91      1010
4      0.92      0.94      0.93      982
5      0.90      0.89      0.90      892
6      0.93      0.96      0.94      958
7      0.94      0.92      0.93      1028
8      0.91      0.89      0.90      974
9      0.91      0.91      0.91      1009

accuracy      0.93      10000
macro avg      0.93      0.93      0.93      10000
weighted avg      0.93      0.93      0.93      10000

```

Confusion Matrix:

```

[[ 961  0  1  3  0  2  9  2  2  0]
 [  0 1110  3  3  0  1  4  2 12  0]
 [ 10  6 935 12 12  3 11 12 25  6]
 [  4  1 14 917  3 31  3  9 15 13]
 [  1  3  4  0 927  0 11  2  4 30]
 [  7  2  3 21  9 795 17  7 22  9]
 [ 10  4  5  1  7 13 915  2  1  0]
 [  2 10 22  8 11  1  1 947  1 25]
 [  9  4  6 19 10 23 14 11 870  8]
 [  9  7  1 12 32 12  1 18  2 915]]

```

313/313 [=====] - 1s 3ms/step

Model with sigmoid activation, categorical\_crossentropy loss, batch size 32, learning rate 0

```

precision      recall      f1-score      support

0      0.95      0.97      0.96      980
1      0.95      0.97      0.96      1135

```

```
2      0.93      0.89      0.91      1032
3      0.92      0.90      0.91      1010
4      0.90      0.93      0.91      982
5      0.88      0.87      0.88      892
6      0.93      0.95      0.94      958
7      0.92      0.92      0.92      1028
8      0.89      0.86      0.88      974
9      0.89      0.89      0.89      1009

accuracy      0.92      0.92      0.92      10000
macro avg     0.92      0.92      0.92      10000
weighted avg  0.92      0.92      0.92      10000

Confusion Matrix:
[[ 955   0   3   1   0   8  10   1   2   0]
 [   0 1106   3   2   1   3   4   1  15   0]
 [   8  10  921  10  12   6  11  15  31   8]
 [   4   2  21  910   2  29   3  15  14  10]
 [   1   5   4   0  913   1  12   1   4  41]
 [   9   4   2  33   9  775  19   8  24   9]
 [   9   5   6   0  11  15  907   2   3   0]
 [   2  14  20   6   7   1   0  943   0  35]
 [   9  14   5  21  14  32  13  15  840  11]
 [  13   6   2  10  45   7   1  23   7  895]]
313/313 [=====] - 1s 4ms/step
Model with sigmoid activation, categorical_crossentropy loss, batch size 32, learning
precision      recall      f1-score      support

0      0.96      0.98      0.97      980
1      0.98      0.99      0.98      1135
2      0.96      0.96      0.96      1032
3      0.94      0.96      0.95      1010
4      0.95      0.96      0.96      982
5      0.96      0.94      0.95      892
6      0.96      0.97      0.97      958
7      0.96      0.95      0.96      1028
8      0.95      0.95      0.95      974
9      0.96      0.93      0.94      1009

accuracy      0.96      0.96      0.96      10000
macro avg     0.96      0.96      0.96      10000
weighted avg  0.96      0.96      0.96      10000

Confusion Matrix:
[[ 964   0   0   1   0   3   9   2   1   0]
 [   0 1121   2   2   0   1   4   1   4   0]
 [   7   3  993   7   3   1   3   9   5   1]
 [   0   1   9  973   1   6   0   9   9   2]
 [   1   0   6   0  944   0   7   1   3  20]
 [   6   2   0  14   4  841  11   1  11   2]
 [   8   3   0   1   3   9  931   0   3   0]
 [   2  11  14   6   4   1   0  978   2  10]
 [   4   1   6  16   6   7   4   8  921   1]
 [   7   5   2  13  24   5   1   5  11  936]]
313/313 [=====] - 2s 4ms/step
Model with sigmoid activation, categorical_crossentropy loss, batch size 32, learning
precision      recall      f1-score      support

0      0.99      0.98      0.98      980
1      0.99      0.99      0.99      1135
2      0.98      0.99      0.98      1032
3      0.96      0.99      0.97      1010
4      0.97      0.98      0.98      982
5      0.98      0.97      0.97      892
6      0.98      0.99      0.98      958
7      0.97      0.98      0.98      1028
8      0.99      0.97      0.98      974
9      0.98      0.96      0.97      1009

accuracy      0.98      0.98      0.98      10000
macro avg     0.98      0.98      0.98      10000
weighted avg  0.98      0.98      0.98      10000

Confusion Matrix:
[[ 962   0   1   2   2   1   4   3   2   3]
 [   0 1127   2   1   0   1   2   1   1   0]
 [   2   2 1017   3   2   0   1   4   1   0]
 [   1   0   4  995   0   2   0   5   3   0]
 [   0   0   1   1  965   0   5   3   0   7]
 [   3   1   0  14   2  862   3   2   2   3]
 [   3   2   1   0   3   4  945   0   0   0]
 [   2   5   7   5   0   0   0 1006   2   1]
 [   1   1   3  10   0   4   3   5  946   1]
 [   0   2   0  10  16   4   2   5   1  969]]
313/313 [=====] - 1s 3ms/step
Model with sigmoid activation, categorical_crossentropy loss, batch size 64, learning
precision      recall      f1-score      support

0      0.93      0.97      0.95      980
1      0.94      0.97      0.96      1135
2      0.92      0.96      0.94      1032
3      0.94      0.97      0.95      1010
4      0.95      0.96      0.95      982
5      0.96      0.95      0.95      892
6      0.96      0.96      0.96      958
7      0.95      0.96      0.95      1028
8      0.97      0.95      0.96      974
9      0.96      0.94      0.95      1009

accuracy      0.95      0.95      0.95      10000
macro avg     0.95      0.95      0.95      10000
weighted avg  0.95      0.95      0.95      10000
```

2	0.93	0.89	0.91	1032
3	0.90	0.90	0.90	1010
4	0.89	0.93	0.91	982
5	0.88	0.86	0.87	892
6	0.93	0.94	0.94	958
7	0.92	0.90	0.91	1028
8	0.88	0.84	0.86	974
9	0.88	0.88	0.88	1009
accuracy			0.91	10000
macro avg	0.91	0.91	0.91	10000
weighted avg	0.91	0.91	0.91	10000

Confusion Matrix:

[	951	0	5	1	0	8	9	2	4	0]
[	0	1103	2	3	0	3	4	1	19	0]
[	15	9	920	10	17	1	11	14	26	9]
[	4	2	20	904	1	30	4	18	17	10]
[	1	6	3	1	915	2	10	1	7	36]
[	11	4	2	36	13	763	18	8	29	8]
[	14	3	5	1	12	15	905	1	2	0]
[	3	20	22	4	9	1	0	929	2	38]
[	10	20	10	30	13	35	12	12	817	15]
[	16	6	3	13	43	8	0	27	7	886]]

313/313 [=====] - 1s 3ms/step  
Model with sigmoid activation, categorical\_crossentropy loss, batch size 64, learning

	precision	recall	f1-score	support
0	0.96	0.98	0.97	980
1	0.97	0.98	0.98	1135
2	0.94	0.93	0.93	1032
3	0.93	0.92	0.93	1010
4	0.93	0.94	0.94	982
5	0.94	0.90	0.92	892
6	0.94	0.96	0.95	958
7	0.95	0.94	0.94	1028
8	0.91	0.93	0.92	974
9	0.92	0.93	0.92	1009
accuracy			0.94	10000
macro avg	0.94	0.94	0.94	10000
weighted avg	0.94	0.94	0.94	10000

Confusion Matrix:

[	959	0	0	2	0	4	9	1	3	2]
[	0	1112	4	2	1	1	4	2	9	0]
[	7	5	956	9	8	2	11	9	20	5]
[	0	3	22	929	1	14	4	14	19	4]
[	1	1	4	1	926	0	10	2	4	33]
[	8	4	5	26	12	800	11	4	17	5]
[	8	3	4	0	7	13	917	2	4	0]
[	1	9	20	4	7	0	0	964	1	22]
[	4	1	5	16	9	15	11	6	901	6]
[	10	7	1	9	25	3	1	10	9	934]]

313/313 [=====] - 1s 3ms/step  
Model with sigmoid activation, categorical\_crossentropy loss, batch size 64, learning

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

Confusion Matrix:

[[	973	0	1	1	1	1	1	1	0]	
[	0	1124	2	3	0	2	2	1	1	0]
[	3	3	1003	4	2	1	5	5	6	0]
[	0	0	1	995	0	1	0	6	3	4]
[	2	0	1	0	961	1	7	1	0	9]
[	5	0	1	14	2	854	7	1	6	2]
[	7	3	1	1	6	4	932	1	3	0]
[	2	12	9	0	1	1	0	995	2	6]
[	7	1	0	5	5	4	3	3	941	5]
[	5	3	1	7	18	2	0	4	4	965]]

313/313 [=====] - 1s 3ms/step  
Model with sigmoid activation, categorical\_crossentropy loss, batch size 128, learning

	precision	recall	f1-score	support
0	0.92	0.97	0.94	980
1	0.92	0.97	0.94	1135
2	0.91	0.86	0.89	1032

```

-
3      0.88      0.88      0.88      1010
4      0.88      0.92      0.90      982
5      0.86      0.82      0.84      892
6      0.91      0.93      0.92      958
7      0.89      0.90      0.90      1028
8      0.87      0.83      0.85      974
9      0.87      0.86      0.87      1009

accuracy
macro avg      0.89      0.89      0.89      10000
weighted avg    0.89      0.89      0.89      10000

```

Confusion Matrix:

```

[[ 946  1  3  2  1 11 11  2  3  0]
 [  0 1101  4  2  1  2  4  2 19  0]
 [ 15 22 889 18 17  0 13 21 30  7]
 [  7  3 18 884  0 43  5 22 19  9]
 [  1  6  4  0 907  1 13  3  6 41]
 [ 16  6  6 51 13 730 22  8 30 10]
 [ 12  4 10  2 15 18 890  3  4  0]
 [  2 19 28  6  9  0  1 924  0 39]
 [ 13 24  9 31 15 32 13 13 804 20]
 [ 12 11  5  9 54  9  1 35  5 868]]

```

313/313 [=====] - 1s 3ms/step

Model with sigmoid activation, categorical\_crossentropy loss, batch size 128, learnir

```

precision recall f1-score support

0      0.95      0.98      0.96      980
1      0.96      0.98      0.97      1135
2      0.94      0.90      0.92      1032
3      0.90      0.91      0.91      1010
4      0.93      0.93      0.93      982
5      0.91      0.87      0.89      892
6      0.93      0.95      0.94      958
7      0.94      0.93      0.93      1028
8      0.90      0.88      0.89      974
9      0.91      0.92      0.91      1009

accuracy
macro avg      0.93      0.93      0.93      10000
weighted avg    0.93      0.93      0.93      10000

```

Confusion Matrix:

```

[[ 961  0  2  1  1  4  9  1  1  0]
 [  0 1113  2  3  1  1  4  2  9  0]
 [ 11  7 925 17  5  4 12 10 33  8]
 [  2  0 19 924  1 23  4 10 18  9]
 [  1  4  3  1 917  0 10  2  5 39]
 [  8  1  2 34  9 780 15 10 25  8]
 [  9  4  4  0  7 15 914  3  2  0]
 [  3 13 17  7  7  2  0 953  1 25]
 [  6 10  6 26 10 28 12 11 858  7]
 [ 11  6  1 13 30  4  0 14  5 925]]

```

313/313 [=====] - 1s 3ms/step

Model with sigmoid activation, categorical\_crossentropy loss, batch size 128, learnir

```

precision recall f1-score support

0      0.97      0.98      0.98      980
1      0.99      0.99      0.99      1135
2      0.97      0.97      0.97      1032
3      0.96      0.98      0.97      1010
4      0.98      0.97      0.97      982
5      0.98      0.95      0.96      892
6      0.97      0.98      0.97      958
7      0.97      0.96      0.96      1028
8      0.95      0.98      0.96      974
9      0.96      0.95      0.96      1009

accuracy
macro avg      0.97      0.97      0.97      10000
weighted avg    0.97      0.97      0.97      10000

```

Confusion Matrix:

```

[[ 963  1  2  2  0  3  5  2  1  1]
 [  0 1122  1  2  0  2  5  1  2  0]
 [  2  2 1000  3  3  0  4  6 12  0]
 [  1  0  4 988  0  1  0  5  7  4]
 [  2  0  4  1 949  0  2  4  2 18]
 [  5  0  2 15  2 849 10  2  5  2]
 [  7  2  0  0  6  4 935  1  3  0]
 [  2  4 11  5  4  0  0 987  6  9]
 [  6  0  2  4  0  3  3  4 951  1]
 [  4  4  1  7  4  6  2  8 10 963]]

```

313/313 [=====] - 1s 3ms/step

Model with sigmoid activation, mean\_squared\_error loss, batch size 32, learning rate

```

precision recall f1-score support

0      0.85      0.95      0.90      980
1      0.82      0.96      0.89      1135
2      0.87      0.78      0.82      1032

```



3	0.85	0.82	0.83	1010
4	0.77	0.85	0.80	982
5	0.80	0.61	0.69	892
6	0.83	0.89	0.86	958
7	0.82	0.85	0.84	1028
8	0.83	0.74	0.78	974
9	0.78	0.75	0.77	1009

accuracy			0.82	10000
macro avg	0.82	0.82	0.82	10000
weighted avg	0.82	0.82	0.82	10000

Confusion Matrix:

[	928	0	4	5	1	12	25	2	2	1]
[	0	1089	5	6	6	3	4	1	20	1]
[	33	35	804	27	21	1	36	25	35	15]
[	10	19	18	825	6	58	11	27	20	16]
[	7	12	7	0	830	2	28	6	6	84]
[	53	40	17	57	48	540	36	44	35	22]
[	23	17	19	1	19	17	855	2	5	0]
[	7	39	25	5	22	0	5	873	7	45]
[	16	51	13	41	38	32	19	16	721	27]
[	16	21	11	8	91	13	9	63	20	757]]

313/313 [=====] - 1s 4ms/step

Model with sigmoid activation, mean\_squared\_error loss, batch size 32, learning rate

	precision	recall	f1-score	support
0	0.93	0.98	0.95	980
1	0.95	0.97	0.96	1135
2	0.92	0.87	0.89	1032
3	0.91	0.90	0.90	1010
4	0.89	0.93	0.91	982
5	0.91	0.83	0.87	892
6	0.92	0.95	0.93	958
7	0.91	0.91	0.91	1028
8	0.87	0.86	0.87	974
9	0.89	0.89	0.89	1009

accuracy			0.91	10000
macro avg	0.91	0.91	0.91	10000
weighted avg	0.91	0.91	0.91	10000

Confusion Matrix:

[	956	0	3	1	0	3	12	2	3	0]
[	0	1102	3	3	1	3	4	2	17	0]
[	15	8	896	14	19	1	10	24	34	11]
[	4	1	22	908	2	22	6	18	19	8]
[	1	5	5	0	912	2	10	1	3	43]
[	11	5	6	39	17	738	22	12	33	9]
[	14	4	4	0	11	11	911	2	1	0]
[	5	20	23	6	8	1	1	931	0	33]
[	11	14	11	23	15	20	17	15	837	11]
[	14	7	2	8	40	10	1	21	12	894]]

313/313 [=====] - 1s 4ms/step

Model with sigmoid activation, mean\_squared\_error loss, batch size 32, learning rate

	precision	recall	f1-score	support
0	0.96	0.98	0.97	980
1	0.97	0.98	0.98	1135
2	0.95	0.91	0.93	1032
3	0.92	0.93	0.92	1010
4	0.92	0.95	0.93	982
5	0.93	0.88	0.91	892
6	0.94	0.95	0.95	958
7	0.93	0.93	0.93	1028
8	0.90	0.92	0.91	974
9	0.92	0.91	0.92	1009

accuracy			0.94	10000
macro avg	0.94	0.94	0.94	10000
weighted avg	0.94	0.94	0.94	10000

Confusion Matrix:

[	962	0	0	2	1	2	8	3	1	1]
[	0	1113	3	2	1	1	4	2	9	0]
[	7	5	943	14	10	3	9	11	26	4]
[	0	0	13	940	4	16	0	10	20	7]
[	1	2	2	1	932	0	9	2	4	29]
[	8	1	1	30	10	786	15	9	23	9]
[	8	4	6	2	8	14	910	3	3	0]
[	2	9	21	4	7	1	0	959	1	24]
[	5	5	3	16	10	14	8	14	896	3]
[	8	6	2	12	31	6	0	13	10	921]]

313/313 [=====] - 1s 3ms/step

Model with sigmoid activation, mean\_squared\_error loss, batch size 64, learning rate

	precision	recall	f1-score	support
0	0.78	0.86	0.82	980
1	0.66	0.97	0.78	1135
2	0.70	0.57	0.62	1032
3	0.72	0.82	0.77	1010
4	0.72	0.82	0.77	982
5	0.72	0.82	0.77	892
6	0.72	0.82	0.77	958
7	0.72	0.82	0.77	1028
8	0.72	0.82	0.77	974
9	0.72	0.82	0.77	1009

3	0.72	0.80	0.75	1010
4	0.71	0.76	0.73	982
5	0.64	0.49	0.55	892
6	0.72	0.81	0.77	958
7	0.78	0.82	0.80	1028
8	0.80	0.39	0.53	974
9	0.68	0.60	0.64	1009
accuracy			0.71	10000
macro avg	0.72	0.71	0.70	10000
weighted avg	0.72	0.71	0.70	10000

Confusion Matrix:

```
[[ 844  4  30  8  2  32  36  15  7  2]
 [  0 1101  9  14  1  3  5  0  0  2]
 [ 31 146 584 53 35  7 104 36  9 27]
 [ 26 33 26 803 2 28 19 29 18 26]
 [ 7 47 6 3 745 14 30 10 7 113]
 [ 84 85 4 129 44 438 32 19 27 30]
 [ 46 37 41 1 16 32 778 3 3 1]
 [ 8 77 23 6 10 3 3 844 1 53]
 [ 29 116 107 84 43 105 41 29 382 38]
 [ 13 31 9 20 152 25 27 98 25 609]]
```

313/313 [=====] - 1s 3ms/step

Model with sigmoid activation, mean\_squared\_error loss, batch size 64, learning rate

	precision	recall	f1-score	support
0	0.91	0.97	0.94	980
1	0.92	0.97	0.95	1135
2	0.91	0.85	0.88	1032
3	0.89	0.89	0.89	1010
4	0.87	0.92	0.90	982
5	0.88	0.79	0.83	892
6	0.90	0.94	0.92	958
7	0.90	0.90	0.90	1028
8	0.87	0.82	0.85	974
9	0.88	0.88	0.88	1009
accuracy			0.89	10000
macro avg	0.89	0.89	0.89	10000
weighted avg	0.89	0.89	0.89	10000

Confusion Matrix:

```
[[ 955  0  1  1  1  4 14  1  3  0]
 [  0 1100  4  3  1  2  5  2 18  0]
 [ 14 18 881 17 17  1 16 23 34 11]
 [ 6 5 27 899 1 29 5 16 15 7]
 [ 0 5 3 0 904 1 15 2 5 47]
 [ 23 11 5 39 29 701 26 15 31 12]
 [ 20 3 5 2 13 12 896 5 2 0]
 [ 3 25 21 6 9 0 0 926 2 36]
 [ 13 18 15 28 16 33 18 17 803 13]
 [ 14 8 6 11 46 10 0 25 6 883]]
```

313/313 [=====] - 1s 3ms/step

Model with sigmoid activation, mean\_squared\_error loss, batch size 64, learning rate

	precision	recall	f1-score	support
0	0.94	0.99	0.96	980
1	0.97	0.98	0.97	1135
2	0.93	0.91	0.92	1032
3	0.93	0.91	0.92	1010
4	0.93	0.94	0.93	982
5	0.92	0.87	0.89	892
6	0.93	0.96	0.95	958
7	0.92	0.92	0.92	1028
8	0.91	0.89	0.90	974
9	0.91	0.91	0.91	1009
accuracy			0.93	10000
macro avg	0.93	0.93	0.93	10000
weighted avg	0.93	0.93	0.93	10000

Confusion Matrix:

```
[[ 966  0  1  1  0  2  7  1  2  0]
 [  0 1110  3  3  1  1  4  2 11  0]
 [ 11 6 939 9 9 2 11 16 22 7]
 [ 3 1 25 919 0 24 3 14 12 9]
 [ 2 3 6 0 920 0 11 2 4 34]
 [ 8 1 2 28 8 779 19 10 29 8]
 [ 11 3 5 0 7 9 919 2 2 0]
 [ 3 8 21 4 9 1 0 950 0 32]
 [ 9 11 6 21 11 23 10 15 865 3]
 [ 15 5 0 7 25 9 0 18 8 922]]
```

313/313 [=====] - 1s 3ms/step

Model with sigmoid activation, mean\_squared\_error loss, batch size 128, learning rate

	precision	recall	f1-score	support
0	0.64	0.85	0.73	980
1	0.61	0.86	0.71	1135
2	0.56	0.37	0.45	1032
3	0.55	0.45	0.50	1010

	0.55	0.47	0.53	982
4	0.63	0.47	0.53	982
5	0.30	0.29	0.29	892
6	0.64	0.70	0.67	958
7	0.61	0.72	0.66	1028
8	0.49	0.40	0.44	974
9	0.46	0.42	0.44	1009
accuracy			0.56	10000
macro avg	0.55	0.55	0.54	10000
weighted avg	0.55	0.56	0.55	10000

Confusion Matrix:  
[[832 11 16 4 20 42 17 22 5 11]  
[ 2 981 26 39 1 48 15 14 5 4]  
[ 70 189 381 19 39 12 162 37 108 15]  
[ 44 100 57 459 13 136 69 43 64 25]  
[ 7 37 67 25 457 88 37 46 43 175]  
[204 65 15 106 17 258 49 25 57 96]  
[ 57 52 34 5 50 31 668 12 45 4]  
[ 4 60 11 7 23 17 6 744 34 122]  
[ 55 92 61 114 8 154 15 48 387 40]  
[ 29 29 11 60 103 75 11 231 41 419]]

313/313 [=====] - 1s 3ms/step  
Model with sigmoid activation, mean\_squared\_error loss, batch size 128, learning rate  
precision recall f1-score support

0	0.90	0.97	0.93	980
1	0.91	0.97	0.94	1135
2	0.90	0.83	0.86	1032
3	0.87	0.86	0.87	1010
4	0.83	0.90	0.87	982
5	0.86	0.75	0.80	892
6	0.90	0.92	0.91	958
7	0.87	0.89	0.88	1028
8	0.86	0.80	0.83	974
9	0.86	0.83	0.85	1009
accuracy			0.88	10000
macro avg	0.88	0.87	0.87	10000
weighted avg	0.88	0.88	0.87	10000

Confusion Matrix:  
[[ 947 0 5 3 0 5 11 2 7 0]  
[ 0 1104 1 4 0 2 4 3 17 0]  
[ 16 19 860 19 22 1 22 34 34 5]  
[ 6 7 22 871 4 39 11 20 21 9]  
[ 2 5 3 1 886 4 16 2 9 54]  
[ 25 15 11 47 42 671 22 22 31 6]  
[ 23 6 9 0 12 17 886 3 2 0]  
[ 5 28 23 4 11 3 3 912 2 37]  
[ 19 28 13 34 23 25 13 14 784 21]  
[ 12 7 12 15 62 17 1 38 6 839]]

313/313 [=====] - 1s 3ms/step  
Model with sigmoid activation, mean\_squared\_error loss, batch size 128, learning rate  
precision recall f1-score support

0	0.93	0.98	0.96	980
1	0.96	0.98	0.97	1135

1

Regression

DATASET03 - AUTO-MPG

macro avg	0.94	0.94	0.94	10000
-----------	------	------	------	-------

- 1. Load and preprocess the given data.
- 2. Build ANN model with regularization and skip connections and train it on the given data.
- 3. Analyze ANN model performance with different batch sizes (test of 3 different batch size) and learning rates (3 different learning rates).
- 4. Plot mse, mae and rmse for different batch size and learning rates.

1

Loading and Preprocessing the data

1

1 data = pd.read\_csv('/content/auto-mpg.csv')

1 data.head(5)

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
0	18.0	8	307.0	130	3504	12.0	70	1	chevrolet chevelle malibu
1	15.0	8	350.0	165	3693	11.5	70	1	buick skylark 320
2	18.0	8	318.0	150	3436	11.0	70	1	plymouth satellite
3	16.0	8	304.0	150	3433	12.0	70	1	amc rebel sst

```
1 data['car name'].nunique()
```

```
305
```

```
1 data.shape
```

```
(398, 9)
```

```
1 data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 398 entries, 0 to 397
Data columns (total 9 columns):
#   Column          Non-Null Count  Dtype
---  -
0   mpg              398 non-null    float64
1   cylinders        398 non-null    int64
2   displacement     398 non-null    float64
3   horsepower       398 non-null    object
4   weight           398 non-null    int64
5   acceleration     398 non-null    float64
6   model year      398 non-null    int64
7   origin           398 non-null    int64
8   car name        398 non-null    object
dtypes: float64(3), int64(4), object(2)
memory usage: 28.1+ KB
```

```
1 # Convert 'horsepower' column to numeric
```

```
2 data['horsepower'] = pd.to_numeric(data['horsepower'], errors='coerce')
```

```
1 data.isnull().sum()
```

```
mpg          0
cylinders    0
displacement 0
horsepower   6
weight       0
acceleration 0
model year   0
origin       0
car name     0
dtype: int64
```

```
1 # Impute missing values with the mean for each column
```

```
2 data['horsepower'] = data['horsepower'].fillna(data['horsepower'].median())
```

```
1 data.isnull().sum()
```

```
mpg          0
cylinders    0
displacement 0
horsepower   0
weight       0
acceleration 0
model year   0
origin       0
car name     0
dtype: int64
```

Car name is of no use

```
1 data = data.drop('car name', axis = 1)
```

```
1 data.shape
```

```
(398, 8)
```

```

1 import pandas as pd
2 import numpy as np
3 from sklearn.model_selection import train_test_split
4 from sklearn.preprocessing import StandardScaler
5
6 # Split the data into features (X) and target (y)
7 X = data.drop(['mpg'], axis=1)
8 y = data['mpg']
9
10 # Split the data into train and test sets
11 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
12
13 # Standardize features
14 scaler = StandardScaler()
15 X_train = scaler.fit_transform(X_train)
16 X_test = scaler.transform(X_test)

```

#### ▼ Build an ANN Model with Regularization and Skip Connections:

```

1 import tensorflow as tf
2 from tensorflow.keras import layers
3
4 def create_model():
5     input_layer = layers.Input(shape=(X_train.shape[1],))
6
7     x = layers.Dense(64, activation='relu', kernel_regularizer=tf.keras.regularizers.l2(0.001))(input_
8     x = layers.BatchNormalization()(x)
9
10    # Add a Dense layer to match input and intermediate layer dimensions
11    identity = layers.Dense(64, activation='linear')(input_layer)
12
13    # Create a skip connection by adding the input to the intermediate layer
14    x = layers.Add()([x, identity])
15
16    x = layers.Dense(64, activation='relu', kernel_regularizer=tf.keras.regularizers.l2(0.001))(x)
17    x = layers.BatchNormalization()(x)
18
19    output_layer = layers.Dense(1)(x) # Output layer for regression
20
21    model = tf.keras.Model(inputs=input_layer, outputs=output_layer)
22
23    return model
24
25 model = create_model()

```

```
1 model
```

```
<keras.src.engine.functional.Functional at 0x7fc01017bc10>
```

#### ▼ Train the ANN Model with Different Batch Sizes and Learning Rates:

```

1 from tensorflow.keras.losses import MeanSquaredError, MeanAbsoluteError
2 from tensorflow.keras.optimizers import Adam
3 from sklearn.metrics import mean_squared_error, mean_absolute_error
4
5 # Define a function to train and evaluate the model with different parameters
6 def train_and_evaluate(batch_size, learning_rate):
7     model = create_model() # Create a new model for each iteration
8     optimizer = Adam(learning_rate=learning_rate)
9     model.compile(optimizer=optimizer, loss='mean_squared_error', metrics=['mae', 'mse'])
10
11    history = model.fit(X_train, y_train, batch_size=batch_size, epochs=100, validation_split=0.2, ver
12
13    y_pred = model.predict(X_test)
14    mse = mean_squared_error(y_test, y_pred)

```

```

15     mae = mean_absolute_error(y_test, y_pred)
16     rmse = np.sqrt(mse)
17
18     return mse, mae, rmse
19
20 # Define lists to store results for different batch sizes and learning rates
21 batch_sizes = [32, 64, 128]
22 learning_rates = [0.001, 0.01, 0.1]
23 results = []
24
25 # Iterate through different batch sizes and learning rates
26 for batch_size in batch_sizes:
27     for learning_rate in learning_rates:
28         mse, mae, rmse = train_and_evaluate(batch_size, learning_rate)
29         results.append((batch_size, learning_rate, mse, mae, rmse))
30
31 # Convert results to a DataFrame for analysis
32 results_df = pd.DataFrame(results, columns=['Batch Size', 'Learning Rate', 'MSE', 'MAE', 'RMSE'])
33
34 3/3 [=====] - 0s 6ms/step
35 3/3 [=====] - 0s 4ms/step
36 3/3 [=====] - 0s 6ms/step
37 WARNING:tensorflow:5 out of the last 13 calls to <function Model.make_predict_function.<locals>.predict_function at 0x7fc00811af80>
38 3/3 [=====] - 0s 8ms/step
39 WARNING:tensorflow:5 out of the last 13 calls to <function Model.make_predict_function.<locals>.predict_function at 0x7fc00811acb0>
40 3/3 [=====] - 0s 5ms/step
41 3/3 [=====] - 0s 5ms/step
42 3/3 [=====] - 0s 5ms/step
43 3/3 [=====] - 0s 5ms/step
44 3/3 [=====] - 0s 4ms/step

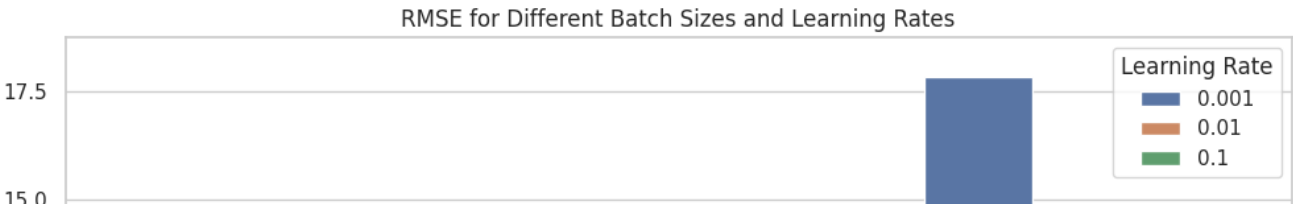
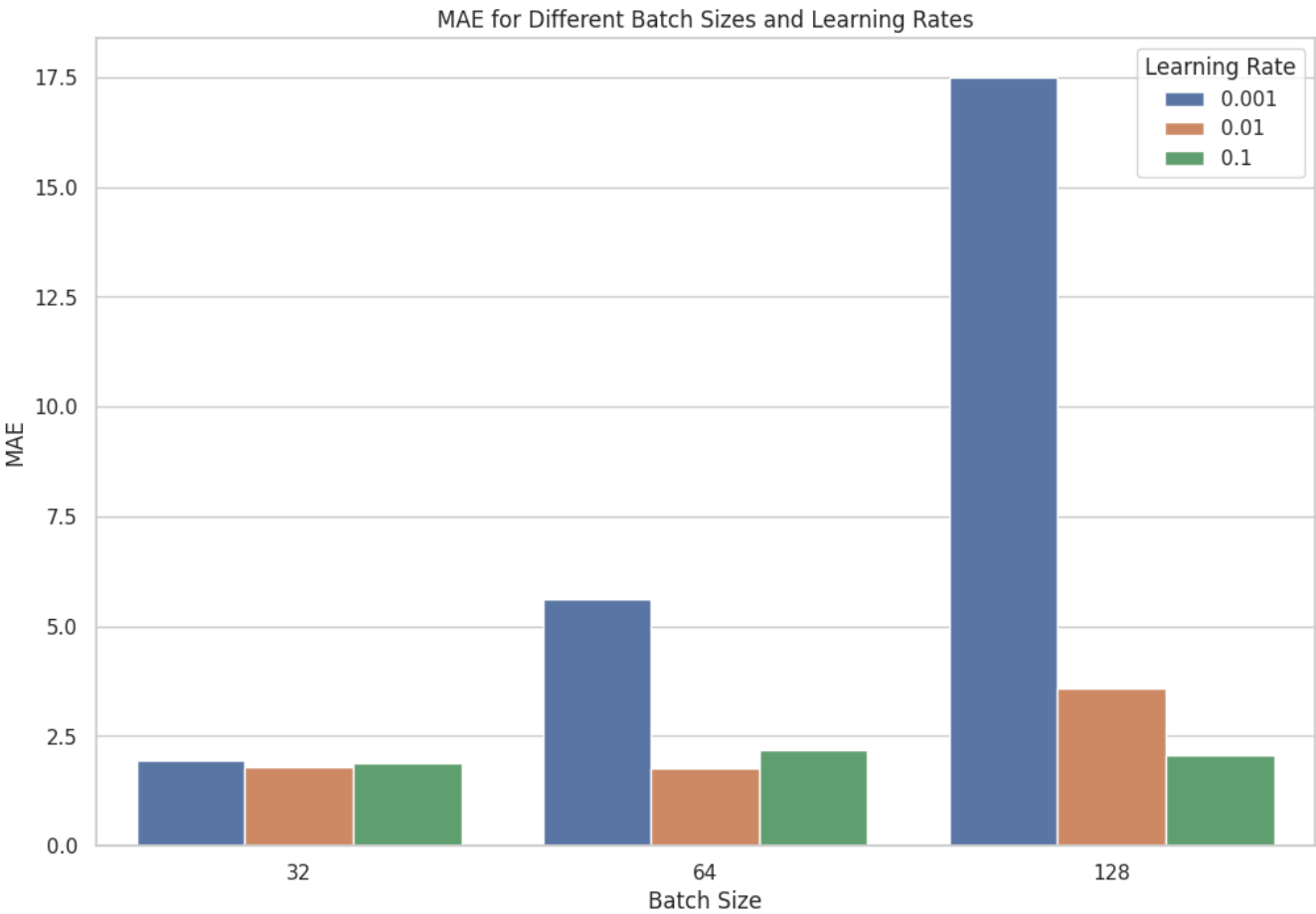
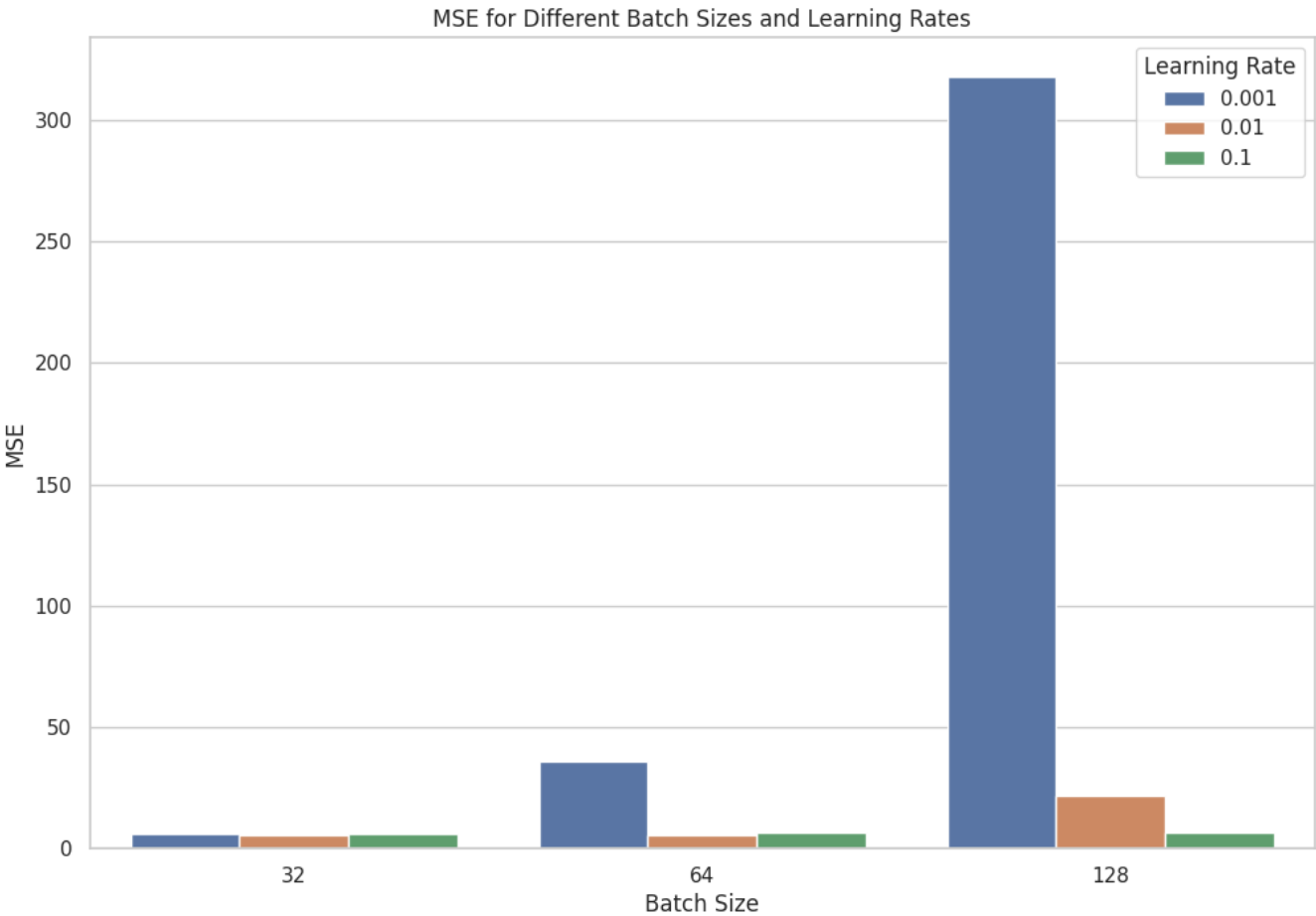
```

#### ▼ Plot MSE, MAE, and RMSE for Different Batch Sizes and Learning Rates:

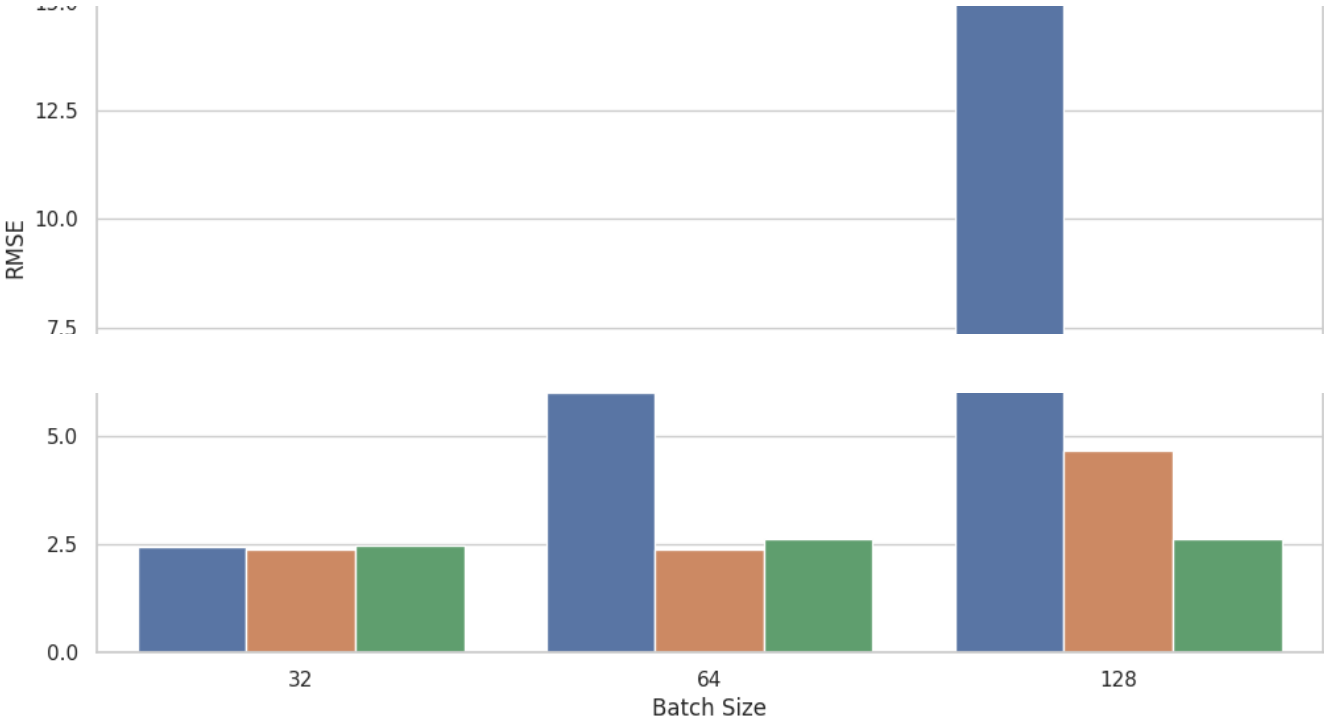
```

1 import matplotlib.pyplot as plt
2 import seaborn as sns
3
4 # Plot MSE, MAE, and RMSE for different batch sizes and learning rates
5 plt.figure(figsize=(12, 8))
6 sns.set(style="whitegrid")
7 sns.barplot(x="Batch Size", y="MSE", hue="Learning Rate", data=results_df)
8 plt.title("MSE for Different Batch Sizes and Learning Rates")
9 plt.show()
10
11 plt.figure(figsize=(12, 8))
12 sns.set(style="whitegrid")
13 sns.barplot(x="Batch Size", y="MAE", hue="Learning Rate", data=results_df)
14 plt.title("MAE for Different Batch Sizes and Learning Rates")
15 plt.show()
16
17 plt.figure(figsize=(12, 8))
18 sns.set(style="whitegrid")
19 sns.barplot(x="Batch Size", y="RMSE", hue="Learning Rate", data=results_df)
20 plt.title("RMSE for Different Batch Sizes and Learning Rates")
21 plt.show()

```



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