# Implementation of simple CNN architecture on MNIST dataset

Hridoy Pal 110146466

#### School of Computer Science

University of Windsor

#### CNN Architecture proposed as follows (Model - 1):

Input Convolution layer (Conv2D), 32 filters, kernel size 5, activation Relu followed by a MaxPooling2D layer, and BatchNormalization.

Convolution layer (Conv2D), 64 filters, kernel size 5, activation Relu followed by a Maxpooling2D layer, and BatchNormalization.

Convolution layer (Conv2D), 128 filters, kernel size 3, activation Relu followed by a MaxPooling2D layer, and BatchNormalization.

Flatten layer.

Dropout probability 0.5

Dense layer with 10 classes.

Implemented the model using Keras API which runs over TensorFlow.

#### **Results Evaluation:**

Training accuracy: 99.58% Validation accuracy: 99.15%

```
469/469 [=
                                        =] - 79s 159ms/step - loss: 0.1567 - accuracy: 0.9528 - val loss: 0.2048 - val accuracy: 0.9435
Epoch 2/10
.
469/469 [==
                                             70s 149ms/step - loss: 0.0527 - accuracy: 0.9843 - val loss: 0.0297 - val accuracy: 0.9904
                                             70s 149ms/step - loss: 0.0368 - accuracy: 0.9883 - val_loss: 0.0354 - val_accuracy: 0.9882
                                             68s 145ms/step - loss: 0.0300 - accuracy: 0.9913 - val_loss: 0.0280 - val_accuracy: 0.9910
.
469/469 [==
Epoch 5/10
.
469/469 [==
                                             71s 152ms/step - loss: 0.0244 - accuracy: 0.9919 - val_loss: 0.0276 - val_accuracy: 0.9914
Epoch 6/10
469/469 [==:
Epoch 7/10
                                             68s 144ms/step - loss: 0.0211 - accuracy: 0.9935 - val_loss: 0.0279 - val_accuracy: 0.9917
469/469 [==
Epoch 8/10
                                             66s 141ms/step - loss: 0.0173 - accuracy: 0.9945 - val loss: 0.0255 - val accuracy: 0.9917
469/469 [==
Epoch 9/10
                                             63s 135ms/step - loss: 0.0160 - accuracy: 0.9949 - val_loss: 0.0238 - val_accuracy: 0.9918
                                             64s 136ms/step - loss: 0.0148 - accuracy: 0.9954 - val_loss: 0.0300 - val_accuracy: 0.9919
.
469/469 [==
Epoch 10/10
 .
69/469 [==:
                                         ] - 65s 139ms/step - loss: 0.0128 - accuracy: 0.9958 - val_loss: 0.0316 - val_accuracy: 0.9915
```

Fig – 1: Training Results

Test accuracy: 99.15%

Fig -2: Test Result

# **Learning Curves:**

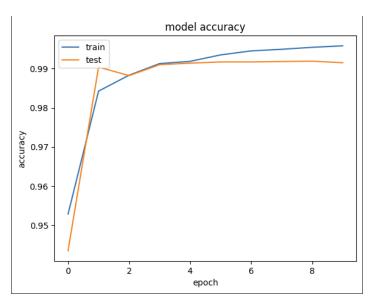


Fig – 3: Model -1 Accuracy

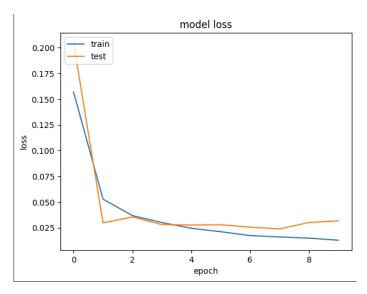


Fig – 4: Model -1 Loss

The model is performing very well on both train and test data with 99.58% and 99.15% accuracy.

# Without BatchNormalization & Dropout (Experimental Model):

Accuracy:

## 0.9904999732971191

Fig – 5: Experimental Model Accuracy

## Learning Curves:

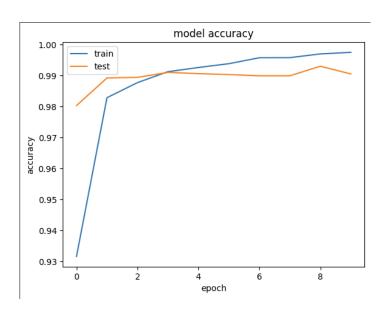


Fig – 6: Experimental Model Accuracy Learning Curve

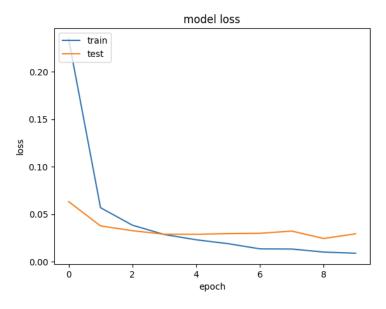


Fig – 7: Experimental Model Loss Curve

Here, the distance between the learning curves for train and test data is bigger compared to Model - 1 (Fig - 3 & Fig - 4). This means that experiment model is overfitting the data. In the experiment model, test accuracy is lower at 99.04%.

# Summary:

By adding BatchNormalization and Dropout (0.5) to Model - 1, data overfitting was slightly reduced, and the accuracy improved to 99.15%. BatchNormalization and Dropout helped to gain higher accuracy and less overfitting.

## **Model Comparison:**

CNN Implementation (Model-1)	Existing CNN Implementation
Accuracy: 99.15	SOPCNN [Yahia Saeed Assiri et al. 2020] Accuracy: 99.83
	Branching/Merging CNN + Homogeneous Vector Capsules [Adam Byerly et al. 2020] Accuracy: 99.87
	Ensemble learning in CNN augmented with fully connected subnetworks [ Daiki Hirata et al. 2020] Accuracy: 9984
	Regularization of Neural Networks using DropConnect [Li Wan et al. 2013] Accuracy: 99.77