



Project Cover Sheet

Assignment Title:	mid_report		
Assignment No:	02	Date of Submission:	Click here to enter a date.
Course Title:	COMPUTER VISION AND PATTERN RECOGNITION		
Course Code:	Click here to enter text.	Section:	A
Semester:	Spring	2019-20	Course Teacher: DEBAJYOTI KARMAKER

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1	Rafshan Bin Razzak	18-38310-2	BSc [CSE]	
2			Choose an item.	
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FACULTY COMMENTS	Marks Obtained	
	Total Marks	

Abstract:

Implement a CNN architecture to classify the MNIST handwritten dataset based and test it with different optimizer (Adam, SGD, RMSProp)

Introduction:

Mnist Handwritten Datasets are test datasets used to train and test purpose in machine learning industry. This is a large datasets where all the characters are handwritten. All the images are grey scaled and the resolution is 28*28 pixels.



CNN or convolutional neural network architecture is a layer based architecture basically used to recognize the written digits. It was created by Mr. Yann LayCun in 1998.

Optimizers determines the way to get the best accuracy by changing the model parameters. In Simple way optimizers works to shape the model in the most accurate way by changing different parameters.

Adam: Gradient descent approaches have a good use in practical life. Adam or adaptive moment estimation is another way of using past gradients to calculate current gradients. This optimizer is practically accepted for use in training neural nets.

SGD: SGD class that implements the stochastic gradient descent optimizer with a learning rate and momentum. First, an instance of the class must be created and configured, then specified to the “optimizer” argument when calling the fit() function on the model.

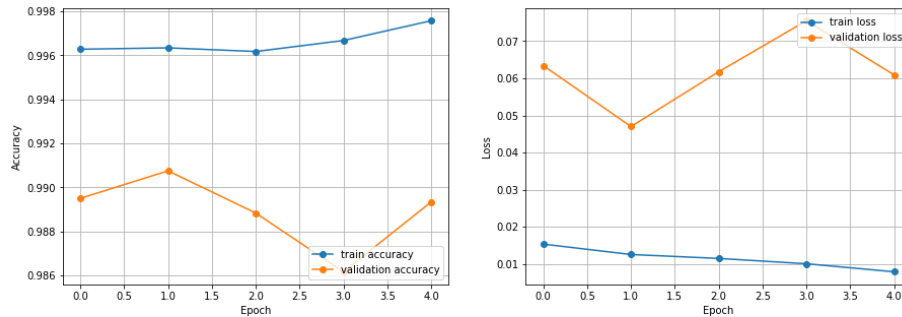
RMSprop :The RMSprop optimizer restricts the oscillations in the vertical direction. Therefore, we can increase our learning rate and our algorithm could take larger steps in the horizontal direction converging faster.

Result:

For Adam Optimizer:

Test Accuracy: 99%

Loss Rate: 0.4%



```
In [62]: test_loss, test_acc = model.evaluate(x_test, y_test)
print('\nTest accuracy:', test_acc)
print('\nTest loss:', test_loss)

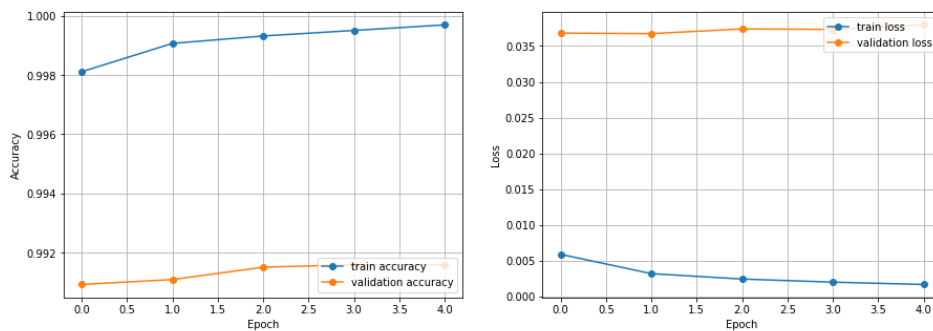
313/313 [=====] - 5s 14ms/step - loss: 0.0457 - accuracy: 0.9911

Test accuracy: 0.99110013256073
Test loss: 0.04568065330386162
```

For SGD Optimizer:

Test Accuracy: 99%

Loss Rate: 0.3%



```
49]: test_loss, test_acc = model.evaluate(x_test, y_test)
print('\nTest accuracy:', test_acc)
print('\nTest loss:', test_loss)

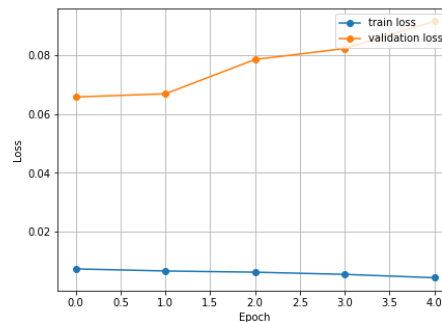
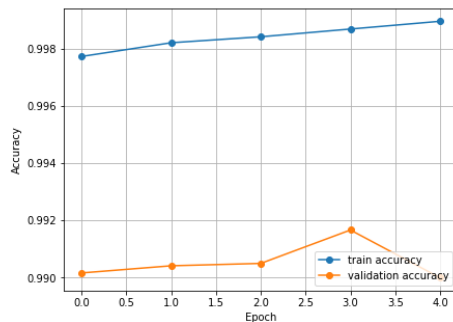
313/313 [=====] - 4s 13ms/step - loss: 0.0311 - accuracy: 0.9920

Test accuracy: 0.9919999837875366
Test loss: 0.031087854877114296
```

For RMSprop Optimizer:

Test Accuracy: 99%

Loss Rate: 0.7%



```
: test_loss, test_acc = model.evaluate(x_test, y_test)
print('\nTest accuracy:', test_acc)
print('\nTest loss:', test_loss)

313/313 [=====] - 4s 13ms/step - loss: 0.0768 - accuracy: 0.9906

Test accuracy: 0.990599898910522

Test loss: 0.0768323689699173
```

Conclusion: The accuracy depends on the model. It varies based on the size, filters and everything. I tried to get accuracy over 98% and I have got over 99% for the given Adam,SGD and RMSprop optimizer. In case of time RMSprop took the least time to calculate. But overall all the optimizers have worked fine and accuracy is nearly 99%.