MULTIMODAL CLASSIFICATION ASSIGNMENT 2

FC Neural Network

1. Hyperparameters

- a. Learning rate (0.01, 0.001)
- b. Optimizer (sgd, adagrad, adam, rmsprop)
- c. Loss
 - Only the sparse categorical cross entropy works as the output is digit and not 1 hot encoding, else we should have used categorical cross entropy.
- d. Activations functions (Sigmoid, Relu, Tanh)
 - i. The activation function for the output layer must be softmax

2. (Using default LR and relu activation function) [Optimizer HP]

Optimizer	SGD	Adagrad	RMSprop	Adam
Train-Acc	0.91	0.91	0.981	0.983
Test-Acc	0.92	0.91	0.975	0.976
Graph	sgd_opt	adagrad_opt	rmsprop_opt	adam_opt

3. (Using relu activation function) [Learning rate HP]

Optimizer	RMSprop		Adam	
LR	0.01	0.01	0.01	0.001
Train-Acc	0.978	0.981	0.975	0.983
Test-Acc	0.967	0.975	0.963	0.974
Graph	rmsprop_0.01	rmsprop_0.001	adam_0.01	adam_0.001

4. (Using Adam and LR = 0.001) [Activation HP]

Activation	Sigmoid	Relu	tanh
Train-Acc	0.966	0.982	0.981
Test-Acc	0.964	0.976	0.973
Graph	sigmoid_act	<u>relu_act</u>	tanh_act

5. Analysis

- a. Best optimizer: Adam and RMSprop with Adam slightly better
- b. Best Learning rate: 0.001 with both Adam and RMSprop
- c. Best Activation Function considering the best optimizer with best LR: Relu slightly better than tanh.

6. Observation

- a. Test accuracy never reached the training accuracy
- b. Overfitting happens some of the time

Convolution Neural Network

1. Hyperparameters

- a. Activation (Relu, Tanh)
- b. Filter size (3, 5, 7)
- c. Stride length (1, 2, 4)
- d. Pool layers (2, 3, 4)

2. Fixing other parameters to see best Activation (3x3, 1, 2 maxpool)

Activation	Relu	Tanh
Train-Acc	0.994	0.991
Test-Acc	0.991	0.986
Graph	<u>Relu_act</u>	<u>Tanh_act</u>

3. Fixing other parameters to see best Filter size and stride (relu, 2 maxpool)

Size	3x3		5x5		7x7				
Stride	1	2	3	1	2	3	1	2	3
Train-Acc	0.994	0.971	0.969	0.994	0.988	0.981	0.991	0.978	0.983
Test-Acc	0.988	0.97	0.965	0.991	0.984	0.98	0.986	0.972	0.981
Graph	fig 1	fig 2	fig 3	fig 4	fig 5	fig 6	fig_7	fig 8	fig 9

Observation:

Decreasing the stride length always result in better accuracy and the opposite happens with the filter size, as size increase the accuracy increases till an overfitting happens

4. Fixing other parameters to se best num of maxpool layers

Maxpool num	2	3	4
Train-Acc	0.994	0.99/0.992	0.985
Test-Acc	0.989	0.988/0.989	0.985

5. Model Analysis

Analysis is done for one Model, however it could be repeated to all similarly

Output	Shape	Param #
(None,	26, 26, 24)	240
(None,	13, 13, 24)	0
(None,	11, 11, 36)	7812
(None,	5, 5, 36)	0
(None,	900)	0
(None,	784)	706384
(None,	128)	100480
(None,	10)	1290
	(None, (None, (None, (None, (None, (None,	Output Shape (None, 26, 26, 24) (None, 13, 13, 24) (None, 11, 11, 36) (None, 5, 5, 36) (None, 900) (None, 784) (None, 128) (None, 10)

Total params: 816,206 Trainable params: 816,206 Non-trainable params: 0

- a. Running time = 26 sec
- b. Number of parameters = 816,206
- c. Number of multiplications
 - Training = 60,000 * ((26*26*24*3*3*1) + (0) + (11*11*36*3*3*24) + (0) + (784*900) + (784*128) + (128*10)) * 5
 - ii. Testing = 10,000 * ((26*26*24*3*3*1) + (0) + (11*11*36*3*3*24) + (0) + (784*900) + (784*128) + (128*10))

6. Comparison to FC NN

Model	FC-NN	CNN
Train Accuracy	0.982	0.994

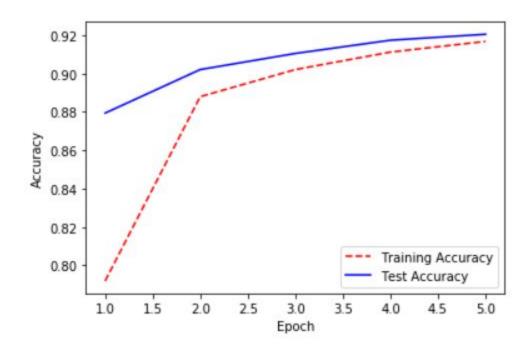
Test Accuracy	0.974	0.991
Time	23.46 sec	26.0 sec

Appendix

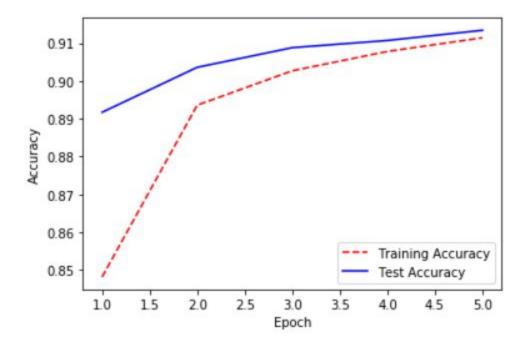
1. FC NN

a. Optimizers

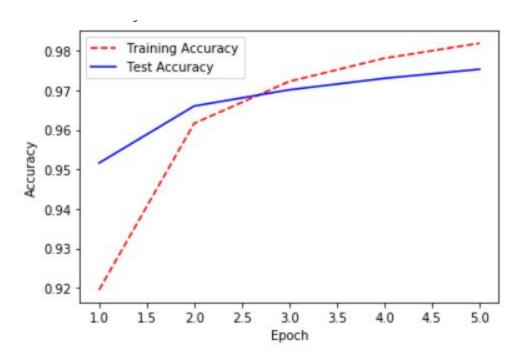
i. <u>Sgd</u>



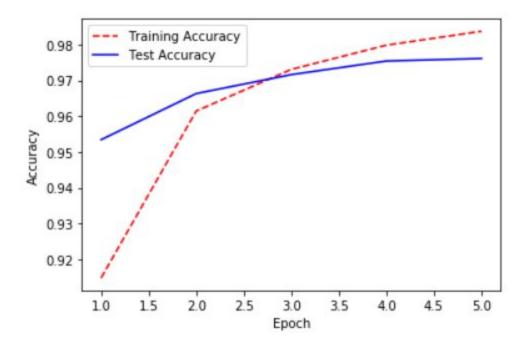
ii. Adagrad



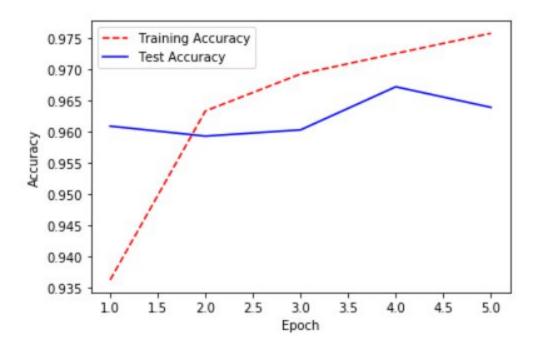
iii. Rmsprop



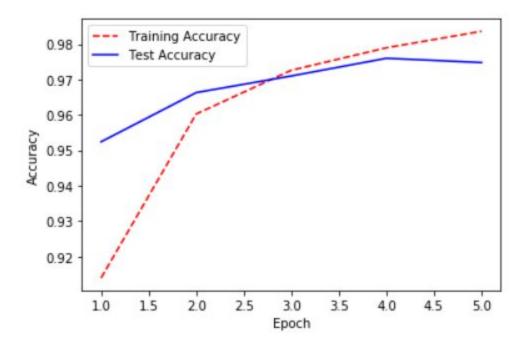
iv. Adam



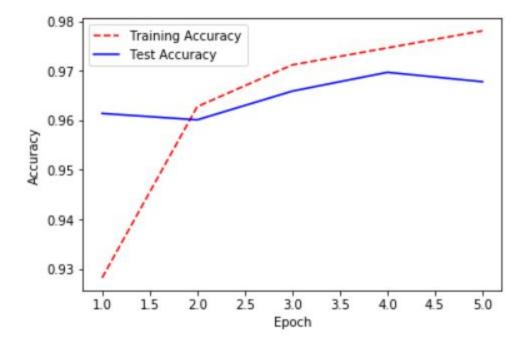
b. LRi. Adam 0.01



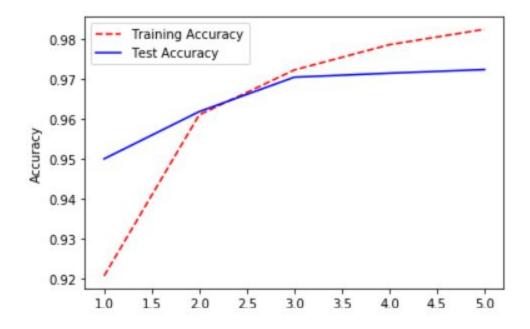
ii. Adam 0.001



iii. RMSprop 0.01

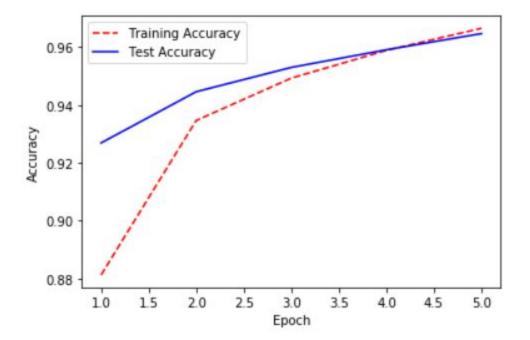


iv. RMSprop 0.001

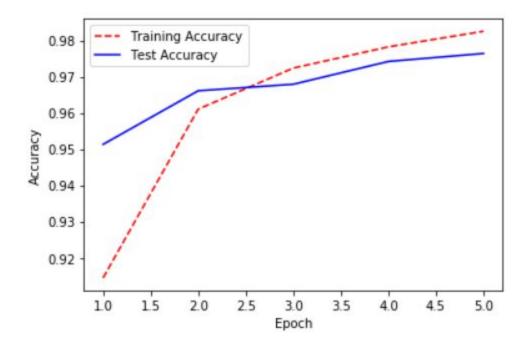


c. Activation

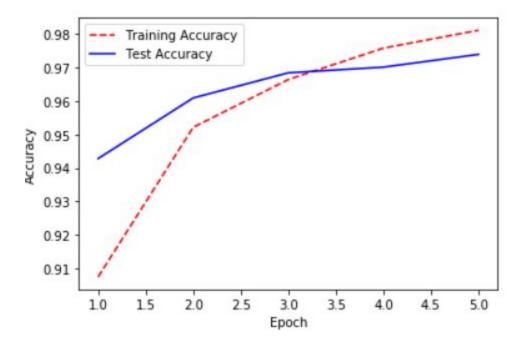
i. Sigmoid



ii. <u>Relu</u>



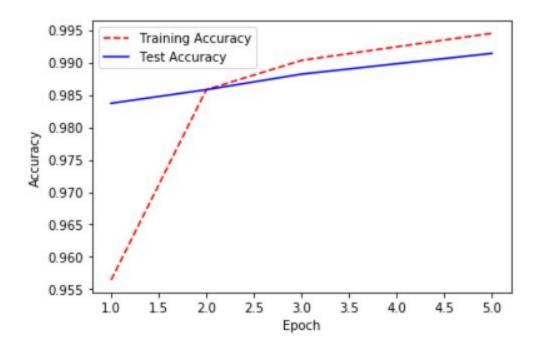
iii. <u>Tanh</u>



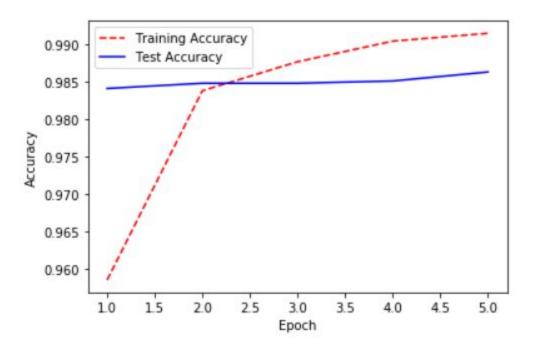
2. CNN

a. Activation

i. Relu

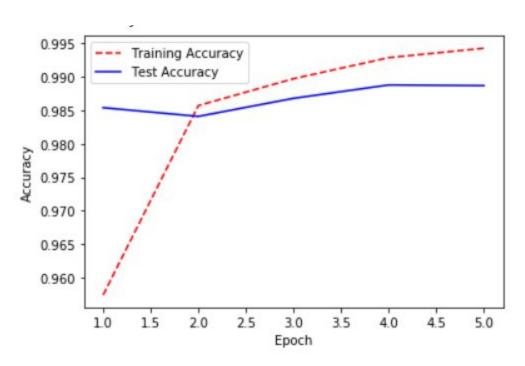


ii. <u>Tanh</u>

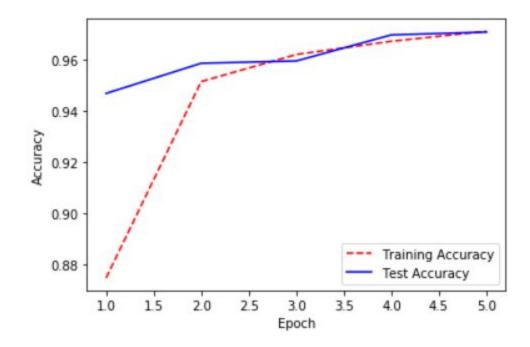


b. Filter size & Stride

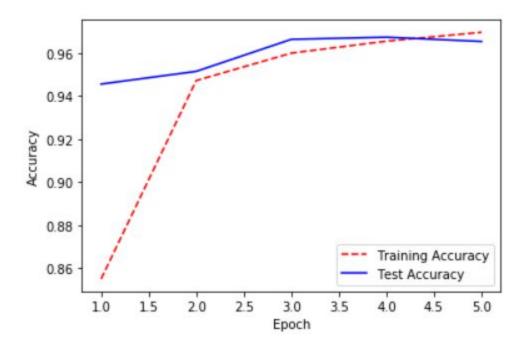
i. <u>3x3 1</u>



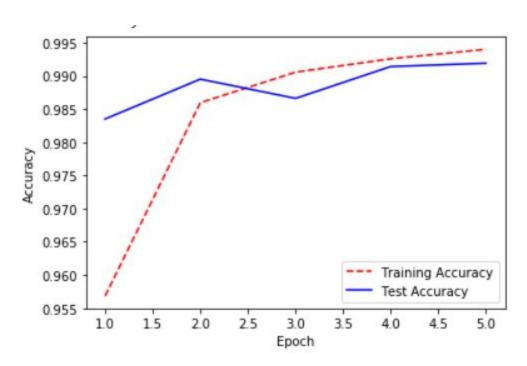
ii. <u>3x3 2</u>



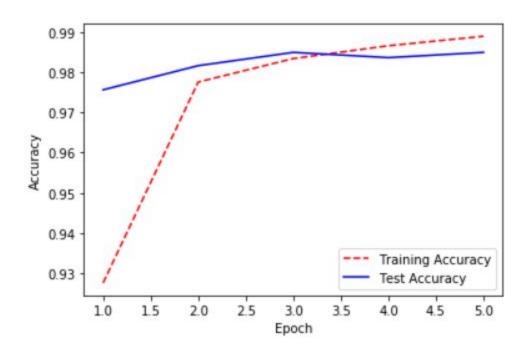
iii. <u>3x3 3</u>



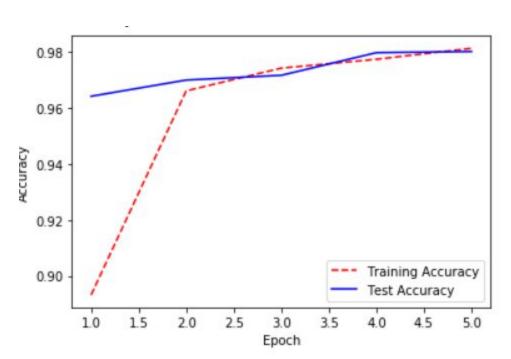
iv. <u>5x5 1</u>



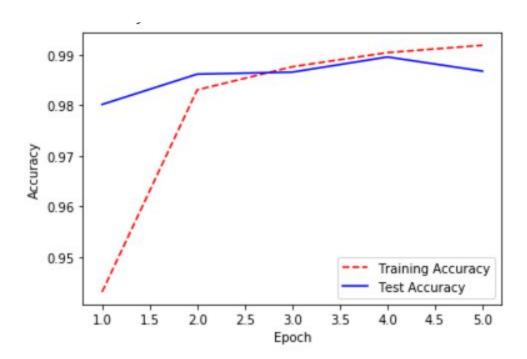




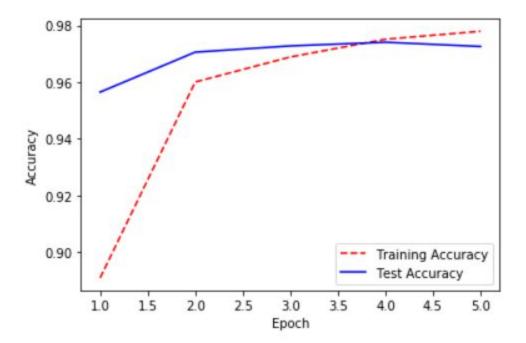
vi. <u>5x5 3</u>



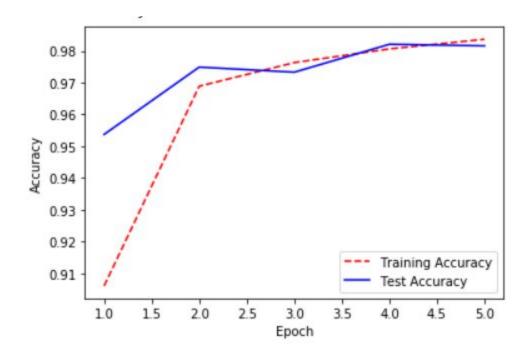
vii. <u>7x7 1</u>



viii. <u>7x7 2</u>



ix. $7x7 \ 3$



c. .