Image Classification Using CNN

CS 6301: Practical Aspects of Data Science

Summer 2020: Assignment 2b

Team Member: -Rajarshi Chattopadhyay (RXC170010)

Introduction

This project involves using Convolutional Neural Network for image classification.

A TensorFlow technique for image classification has been used in with the help of the Keras API of TensorFlow in Python.

Data

CIFAR-10 - Object Recognition in Images http://www.cs.utoronto.ca/~kriz/cifar.html The data is also stored in https://www.utdallas.edu/~rxc170010/cifar-10-batches-py/

The CIFAR-10 is labeled subset of the <u>80 million tiny images</u> dataset, consisting of 60000 32x32 color images in 10 classes, with 6000 images per class.

The data, which comes as vectors, has been reshaped to 32x32x3, indicating a 32x32 image with 3 channels for RGB. This is necessary for using as input for the 2D Convolutional layers of the model.

Training examples 45000, Testing examples 5000

Models

Models were created with different architectures

- The number of neurons in each kernel and the number of hidden layers were modified to test and improve the model.
- Row and column size from input to hidden layer has been kept the same.
- Batch normalization was used in the input layer to ensure there is not much covariance shift from its output.
- Max-pooling was used in the hidden Conv layer to reduce row and column dimensions to half.
- Dropout was used to inactivate randomly 50% of the nodes in Conv layers and 40% in Dense layers for each data example at the beginning. Then changed later.

Library used: Keras https://www.tensorflow.org/guide/keras/sequential model

Training

The models were trained and tuned to find the best classifier

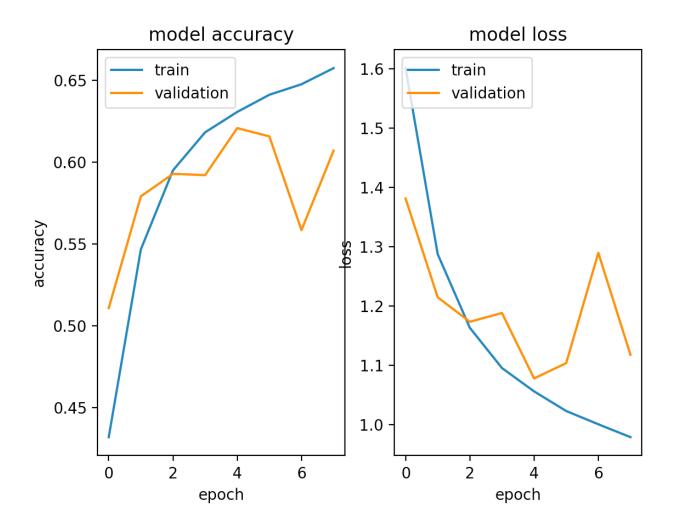
- Validation was done during training with 33% of the training data
- Batch size of 32 was used
- Number of epochs were changed to improve the model
- Early stopping was done to stop training if there is no significant improvement in validation accuracy after every 3 epochs.
- Checkpointing has been performed to save the best performing model for later use during prediction.

Testing

Each model was tested with test data to obtain the test accuracy and loss. Model: "sequential_1"

Layer (type)	Output Shape	Param	n #	
conv2d_1 (Conv2D)	(None, 32, 32,	32) 8	896	
batch_normalization_	1 (Batch (None, 32, 3	2, 32)	128	
conv2d_2 (Conv2D)	(None, 30, 30,	16) 4	4624	
max_pooling2d_1 (M	axPooling2 (None, 15	5, 15, 16)) 0	
dropout_1 (Dropout)	(None, 15, 15, 1	6) 0)	
flatten_1 (Flatten)	(None, 3600)	0		
dense_1 (Dense)	(None, 10)	36010	0	
Total params: 41,658 Trainable params: 41, Non-trainable params				
Train on 30149 sample Epoch 1/10 30149/30149 [===== 0.4320 - val_loss: 1.3 Epoch 2/10	· 			ss: 1.6011 - accuracy:
30149/30149 [===== 0.5468 - val_loss: 1.2 Epoch 3/10	145 - val_accuracy: 0	.5792		ss: 1.2873 - accuracy:
0.5951 - val_loss: 1.1 Epoch 4/10	732 - val_accuracy: 0	.5930	====] - 53s 2ms/step - lo	·
30149/30149 [===== 0.6183 - val_loss: 1.1 Epoch 5/10	880 - val_accuracy: 0	.5921		ss: 1.0952 - accuracy:
0.6308 - val_loss: 1.0 Epoch 6/10			====] - 51s 2ms/step - lo	ss: 1.0558 - accuracy:
30149/30149 [===== 0.6413 - val_loss: 1.10 Epoch 7/10			====] - 52s 2ms/step - lo	ss: 1.0227 - accuracy:
			====] - 51s 2ms/step - lo	ss: 1.0003 - accuracy:

Train result: train_loss: 0.9786 - train_accuracy: 65.7567
 Validation result: val_loss: 1.1175 - val_accuracy: 60.7165
 Test result: test_loss: 1.0774 - test_accuracy: 62.1800

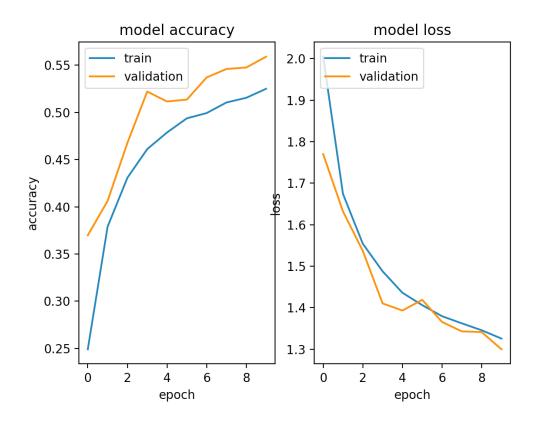


Model: "sequential_2"

Layer (type)	Output Shape	Param #	
conv2d_1 (Conv2D)	(None, 32, 32, 3	32) 896	
batch_normalization_	1 (Batch (None, 32, 32	2, 32) 128	
conv2d_2 (Conv2D)	(None, 30, 30, 1	6) 4624	
max_pooling2d_1 (M	axPooling2 (None, 15	, 15, 16) 0	
dropout_1 (Dropout)	(None, 15, 15, 16	6) 0	
conv2d_3 (Conv2D)	(None, 13, 13, 8	3) 1160	
max_pooling2d_2 (M	axPooling2 (None, 6,	6, 8) 0	
dropout_2 (Dropout)	(None, 6, 6, 8)	0	
flatten_1 (Flatten)	(None, 288)	0	
dense_1 (Dense)	(None, 10)	2890	
Total params: 9,698 Trainable params: 9,6 Non-trainable params			
Epoch 1/10	les, validate on 14851	_	0040
0.2490 - val_loss: 1.7	======================================	=======] - 63s 2ms/step - loss: 2 3698	.0049 - accuracy:
_	======================================		.6745 - accuracy:
	======================================	=======] - 57s 2ms/step - loss: 1 4678	.5544 - accuracy:
0.4611 - val_loss: 1.4 Epoch 5/10	105 - val_accuracy: 0.		·
	933 - val_accuracy: 0.	=======] - 60s 2ms/step - loss: 1 5116	.4362 - accuracy:

```
30149/30149 [=
                                                        =] - 59s 2ms/step - loss: 1.4063 - accuracy:
0.4938 - val loss: 1.4191 - val accuracy: 0.5136
Epoch 7/10
30149/30149 [==
                                                        =] - 61s 2ms/step - loss: 1.3797 - accuracy:
0.4994 - val loss: 1.3658 - val accuracy: 0.5371
Epoch 8/10
30149/30149 [==
                                                        =] - 60s 2ms/step - loss: 1.3624 - accuracy:
0.5106 - val loss: 1.3430 - val accuracy: 0.5461
Epoch 9/10
30149/30149 [=
                                                        =] - 72s 2ms/step - loss: 1.3458 - accuracy:
0.5155 - val loss: 1.3414 - val accuracy: 0.5476
Epoch 10/10
30149/30149 [=
                                                        =] - 66s 2ms/step - loss: 1.3257 - accuracy:
0.5251 - val loss: 1.3000 - val accuracy: 0.5591
5000/5000 [=
                                                      -] - 2s 461us/step
```

- Train result: train_loss: 1.3257 train_accuracy: 52.5092
 Validation result: val_loss: 1.3000 val_accuracy: 55.9087
- o Test result: test_loss: 1.2767 test_accuracy: 56.8000



Model: "sequential_3"

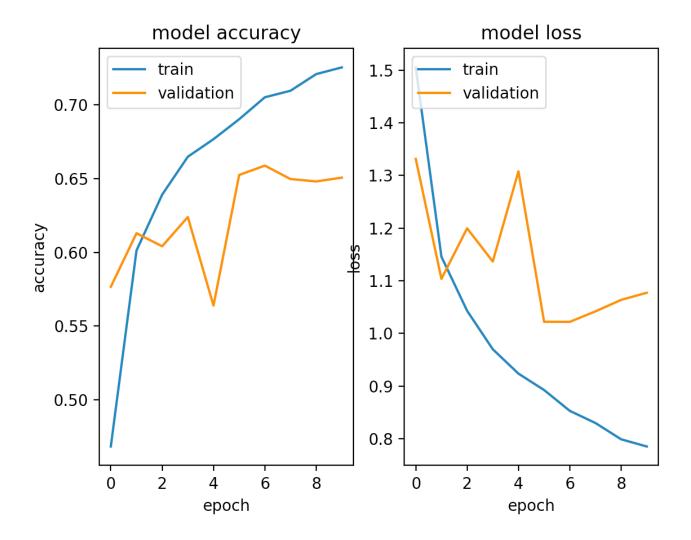
Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)	(None, 32, 32, 6	64) 1792
batch_normalization_	1 (Batch (None, 32, 32)	32, 64) 256
conv2d_2 (Conv2D)	(None, 30, 30, 3	32) 18464
max_pooling2d_1 (M	TaxPooling2 (None, 15	5, 15, 32) 0
dropout_1 (Dropout)	(None, 15, 15, 32	32) 0
conv2d_3 (Conv2D)	(None, 13, 13, 8	8) 2312
max_pooling2d_2 (M	TaxPooling2 (None, 6,	, 6, 8) 0
dropout_2 (Dropout)	(None, 6, 6, 8)	0
flatten_1 (Flatten)	(None, 288)	0
dense_1 (Dense)	(None, 10)	2890
Total params: 25,714 Trainable params: 25, Non-trainable params	,586	
Epoch 1/20 30149/30149 [=====		======] - 93s 3ms/step - loss: 2.0687 - accuracy
Epoch 2/20 30149/30149 [===== 0.2337 - val_loss: 1.8	164 - val_accuracy: 0. 974 - val_accuracy: 0.	======] - 62s 2ms/step - loss: 2.0563 - accuracy
Epoch 3/20 30149/30149 [===== 0.3720 - val_loss: 1.6	 212 - val_accuracy: 0.	=======] - 57s 2ms/step - loss: 1.6948 - accuracy

Not much accuracy increase, so forcefully interrupted training.

Model: "sequential_4"

Layer (type)	Output Shape	Param #	!	
conv2d_1 (Conv2D)	(None, 32, 32,	64) 17	======================================	
batch_normalization_	1 (Batch (None, 32, 3	32, 64)	256	
conv2d_2 (Conv2D)	(None, 30, 30,	32) 18	3464	
max_pooling2d_1 (M	IaxPooling2 (None, 1	5, 15, 32)	0	
dropout_1 (Dropout)	(None, 15, 15, 3	32) 0		
flatten_1 (Flatten)	(None, 7200)	0		
dense_1 (Dense)	(None, 10)	72010		
Total params: 92,522 Trainable params: 92 Non-trainable params	,394			
0.6390 - val_loss: 1.1 Epoch 4/20 30149/30149 [===== 0.6647 - val_loss: 1.1 Epoch 5/20 30149/30149 [===== 0.6766 - val_loss: 1.3	315 - val_accuracy: (035 - val_accuracy: (997 - val_accuracy: (366 - val_accuracy: (0.6129	===] - 95s 3ms/step - los ===] - 92s 3ms/step - los ===] - 91s 3ms/step - los ===] - 90s 3ms/step - los ===] - 103s 3ms/step - los	s: 1.1457 - accuracy: s: 1.0429 - accuracy: s: 0.9699 - accuracy:
0.6901 - val_loss: 1.0 Epoch 7/20	221 - val_accuracy: (0.6523	===] - 100s 3ms/step - los	·

Train result: train_loss: 0.7851 - train_accuracy: 72.5165
Validation result: val_loss: 1.0772 - val_accuracy: 65.0529
Test result: test_loss: 1.0292 - test_accuracy: 67.7200

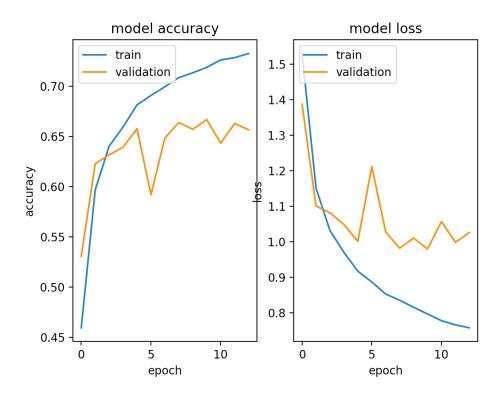


Model: "sequential_5"

Layer (type)	Output Shape	Par	ram #	
conv2d_1 (Conv2I	(None, 32, 32	2, 16)	448	
batch_normalization	on_1 (Batch (None, 32	, 32, 16) 64	
conv2d_2 (Conv2I	(None, 30, 30	0, 32)	4640	
max_pooling2d_1	(MaxPooling2 (None,	15, 15,	32)	0
dropout_1 (Dropou	(None, 15, 15	, 32)	0	
flatten_1 (Flatten)	(None, 7200)	0		
dense_1 (Dense)	(None, 10)	72	2010	
Total params: 77,1 Trainable params: Non-trainable para	77,130			
	mples, validate on 148	51 samp	oles	
Epoch 1/20 30149/30149 [===] - 48s 2ms/step - loss: 1.5301 - accura
_	1.3875 - val accuracy:	0.5305	-	1 403 21113/3tep 1033. 1.3301 decura
Epoch 2/20	_ •			
30149/30149 [===		0.6226	-] - 44s 1ms/step - loss: 1.1513 - accura
0.5968 - val_loss: Epoch 3/20	1.1007 - val_accuracy:	0.6228	ì	
30149/30149 [===	=======================================] - 42s 1ms/step - loss: 1.0318 - accura
-	1.0817 - val accuracy:	0.6317	_	, . <u>-</u> 2 1
Epoch 4/20	_ •			
_			_] - 42s 1ms/step - loss: 0.9705 - accura
	1.0489 - val_accuracy:	0.6392	•	
Epoch 5/20] - 52s 2ms/step - loss: 0.9173 - accura
_	1.0012 - val accuracy:		_	j - 328 21118/step - 1088. 0.91/3 - accura
Epoch 6/20	1.0012 var_accuracy.	0.0570	,	
			7] - 42s 1ms/step - loss: 0.8869 - accura
	1.2114 - val accuracy:			
Epoch 7/20	_ •			
30149/30149 [===]] - 48s 2ms/step - loss: 0.8529 - accura
_	1.0276 - val_accuracy:	0.6486	;)	
Epoch 8/20				
-			_] - 48s 2ms/step - loss: 0.8353 - accura
0.7086 - val loss:	0.9819 - val accuracy:	- () 6639)	
Epoch 9/20	o.yory var_accuracy.	. 0.0057	•	

```
30149/30149 [=
                                                        =] - 44s 1ms/step - loss: 0.8156 - accuracy:
0.7134 - val loss: 1.0106 - val accuracy: 0.6571
Epoch 10/20
30149/30149 [==
                                                        =] - 45s 1ms/step - loss: 0.7968 - accuracy:
0.7188 - val loss: 0.9798 - val accuracy: 0.6668
Epoch 11/20
30149/30149 [==
                                                        =] - 44s 1ms/step - loss: 0.7779 - accuracy:
0.7263 - val loss: 1.0569 - val accuracy: 0.6435
Epoch 12/20
30149/30149 [==
                                                        =] - 44s 1ms/step - loss: 0.7662 - accuracy:
0.7285 - val loss: 0.9988 - val accuracy: 0.6630
Epoch 13/20
                                                        -] - 47s 2ms/step - loss: 0.7578 - accuracy:
30149/30149 [=
0.7326 - val loss: 1.0258 - val accuracy: 0.6567
5000/5000 [=
                                                      -] - 2s 375us/step
```

- Train result: train_loss: 0.7578 train_accuracy: 73.2628
 Validation result: val loss: 1.0258 val accuracy: 65.6656
- o Test result: test loss: 1.0088 test accuracy: 67.2400

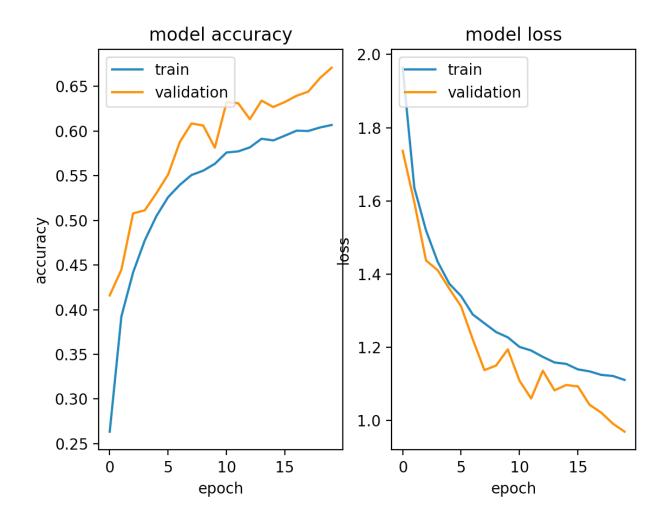


Model: "sequential_6"

Layer (type)	Output Shape	Param #	
conv2d_1 (Conv2D)	(None, 32, 32,	16) 448	
batch_normalization_1	l (Batch (None, 32, 3	2, 16) 64	
conv2d_2 (Conv2D)	(None, 30, 30,	32) 4640	
max_pooling2d_1 (Ma	axPooling2 (None, 15	5, 15, 32) 0	
dropout_1 (Dropout)	(None, 15, 15, 3	2) 0	
conv2d_3 (Conv2D)	(None, 13, 13,	16) 4624	
max_pooling2d_2 (Ma	axPooling2 (None, 6,	6, 16) 0	
dropout_2 (Dropout)	(None, 6, 6, 16)	0	
flatten_1 (Flatten)	(None, 576)	0	
dense_1 (Dense)	(None, 100)	57700	
dropout_3 (Dropout)	(None, 100)	0	
dense_2 (Dense)	(None, 10)	1010	
Total params: 68,486 Trainable params: 68,4 Non-trainable params:			
Train on 30149 sample Epoch 1/20	es, validate on 14851	samples	
30149/30149 [====== 0.2634 - val_loss: 1.73			loss: 1.9638 - accuracy:
0.3926 - val_loss: 1.59			loss: 1.6356 - accuracy:
Epoch 3/20 30149/30149 [====== 0.4417 - val loss: 1.43		=======] - 48s 2ms/step -	loss: 1.5193 - accuracy:
Epoch 4/20 ====================================	_	=====] - 49s 2ms/step -	loss: 1.4334 - accuracy:
0.4775 - val_loss: 1.41 Epoch 5/20	05 - val_accuracy: 0	.5113	

```
30149/30149 [======
                                               =] - 48s 2ms/step - loss: 1.3734 - accuracy:
0.5049 - val loss: 1.3598 - val accuracy: 0.5305
Epoch 6/20
=] - 57s 2ms/step - loss: 1.3400 - accuracy:
0.5260 - val loss: 1.3120 - val accuracy: 0.5513
Epoch 7/20
30149/30149 [======
                                               =] - 45s 1ms/step - loss: 1.2897 - accuracy:
0.5397 - val loss: 1.2218 - val accuracy: 0.5877
Epoch 8/20
30149/30149 [======
                                               =] - 44s 1ms/step - loss: 1.2656 - accuracy:
0.5507 - val loss: 1.1377 - val accuracy: 0.6084
Epoch 9/20
=] - 47s 2ms/step - loss: 1.2416 - accuracy:
0.5556 - val loss: 1.1500 - val accuracy: 0.6062
Epoch 10/20
=] - 44s 1ms/step - loss: 1.2273 - accuracy:
0.5633 - val loss: 1.1943 - val accuracy: 0.5814
Epoch 11/20
30149/30149 [=
                                               =] - 44s 1ms/step - loss: 1.2011 - accuracy:
0.5760 - val loss: 1.1089 - val accuracy: 0.6322
Epoch 12/20
30149/30149 [=
                                               =] - 45s 2ms/step - loss: 1.1911 - accuracy:
0.5772 - val loss: 1.0602 - val accuracy: 0.6311
Epoch 13/20
30149/30149 [============
                                               = ] - 46s 2ms/step - loss: 1.1738 - accuracy:
0.5817 - val loss: 1.1360 - val accuracy: 0.6133
Epoch 14/20
30149/30149 [=========
                                               =] - 49s 2ms/step - loss: 1.1586 - accuracy:
0.5914 - val loss: 1.0824 - val accuracy: 0.6341
Epoch 15/20
=] - 47s 2ms/step - loss: 1.1547 - accuracy:
0.5895 - val loss: 1.0969 - val accuracy: 0.6269
Epoch 16/20
= 1 - 47s 2ms/step - loss: 1.1397 - accuracy:
0.5948 - val loss: 1.0932 - val accuracy: 0.6326
Epoch 17/20
30149/30149 [=====
                                               =] - 45s 1ms/step - loss: 1.1342 - accuracy:
0.6004 - val loss: 1.0431 - val accuracy: 0.6395
Epoch 18/20
30149/30149 [======
                                               =] - 45s 1ms/step - loss: 1.1246 - accuracy:
0.6001 - val loss: 1.0215 - val accuracy: 0.6441
Epoch 19/20
30149/30149 [========
                                             ===] - 47s 2ms/step - loss: 1.1217 - accuracy:
0.6040 - val loss: 0.9914 - val accuracy: 0.6593
Epoch 20/20
```

- Train result: train_loss: 1.1109 train_accuracy: 60.6786
 Validation result: val_loss: 0.9694 val_accuracy: 67.0931
- o Test result: test_loss: 0.9389 test_accuracy: 68.8400

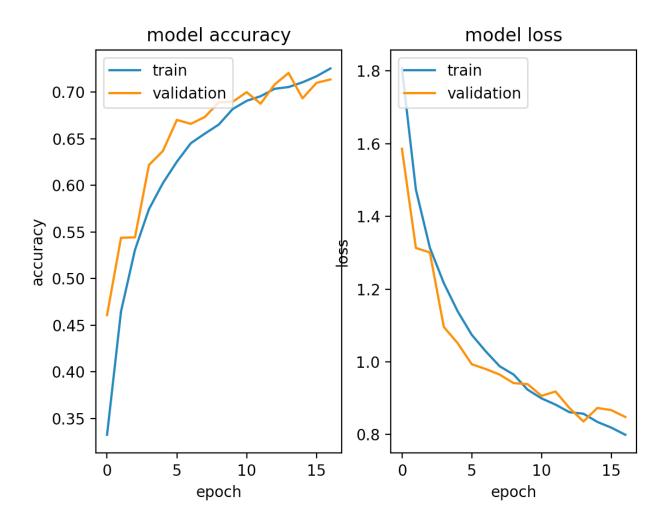


Model: "sequential_7"

Layer (type)	Output Shape	Param #	
conv2d_1 (Conv2D)	(None, 32, 32,	16) 448	
batch_normalization_	1 (Batch (None, 32, 3	2, 16) 64	
conv2d_2 (Conv2D)	(None, 30, 30, 3	32) 4640	
max_pooling2d_1 (M	axPooling2 (None, 15	5, 15, 32) 0	
dropout_1 (Dropout)	(None, 15, 15, 3	2) 0	
conv2d_3 (Conv2D)	(None, 13, 13,	64) 18496	
max_pooling2d_2 (M	axPooling2 (None, 6,	6, 64) 0	
dropout_2 (Dropout)	(None, 6, 6, 64)	0	
flatten_1 (Flatten)	(None, 2304)	0	
dense_1 (Dense)	(None, 100)	230500	
dropout_3 (Dropout)	(None, 100)	0	
dense_2 (Dense)	(None, 100)	10100	
dropout_4 (Dropout)	(None, 100)	0	
dense_3 (Dense)	(None, 10)	1010	
Total params: 265,258 Trainable params: 265 Non-trainable params	5,226		
0.3323 - val_loss: 1.5 Epoch 2/20 30149/30149 [===== 0.4647 - val_loss: 1.3 Epoch 3/20	862 - val_accuracy: 0		s: 1.4738 - accuracy:
0.5307 - val_loss: 1.3			s. 1.5159 - accuracy:

```
Epoch 4/20
30149/30149 [==
                                              =] - 53s 2ms/step - loss: 1.2165 - accuracy:
0.5743 - val loss: 1.0960 - val accuracy: 0.6217
Epoch 5/20
30149/30149 [=============
                                              =] - 74s 2ms/step - loss: 1.1384 - accuracy:
0.6023 - val loss: 1.0509 - val accuracy: 0.6368
Epoch 6/20
=] - 56s 2ms/step - loss: 1.0744 - accuracy:
0.6253 - val loss: 0.9934 - val accuracy: 0.6701
Epoch 7/20
30149/30149 [======
                                              =] - 54s 2ms/step - loss: 1.0289 - accuracy:
0.6451 - val loss: 0.9802 - val accuracy: 0.6659
Epoch 8/20
= ] - 52s 2ms/step - loss: 0.9878 - accuracy:
0.6554 - val loss: 0.9649 - val accuracy: 0.6732
Epoch 9/20
30149/30149 [======
                                              =] - 53s 2ms/step - loss: 0.9650 - accuracy:
0.6651 - val loss: 0.9414 - val accuracy: 0.6890
Epoch 10/20
30149/30149 [=====
                                               =] - 54s 2ms/step - loss: 0.9234 - accuracy:
0.6818 - val loss: 0.9386 - val accuracy: 0.6896
Epoch 11/20
30149/30149 [=======
                                              =] - 51s 2ms/step - loss: 0.8994 - accuracy:
0.6906 - val loss: 0.9064 - val accuracy: 0.6999
Epoch 12/20
= ] - 59s 2ms/step - loss: 0.8823 - accuracy:
0.6955 - val loss: 0.9184 - val accuracy: 0.6875
Epoch 13/20
30149/30149 [======
                                              =] - 51s 2ms/step - loss: 0.8614 - accuracy:
0.7035 - val loss: 0.8729 - val accuracy: 0.7081
Epoch 14/20
30149/30149 [============
                                              = ] - 51s 2ms/step - loss: 0.8573 - accuracy:
0.7054 - val loss: 0.8359 - val accuracy: 0.7205
Epoch 15/20
30149/30149 [===========
                                              =] - 52s 2ms/step - loss: 0.8347 - accuracy:
0.7105 - val loss: 0.8732 - val accuracy: 0.6933
Epoch 16/20
30149/30149 [======
                                               =] - 56s 2ms/step - loss: 0.8189 - accuracy:
0.7169 - val loss: 0.8671 - val accuracy: 0.7099
Epoch 17/20
==] - 52s 2ms/step - loss: 0.7993 - accuracy:
0.7253 - val loss: 0.8482 - val accuracy: 0.7136
5000/5000 [=====] - 2s 400us/step
```

- o Train result: train loss: 0.7993 train accuracy: 72.5331
- O Validation result: val loss: 0.8482 val accuracy: 71.3555

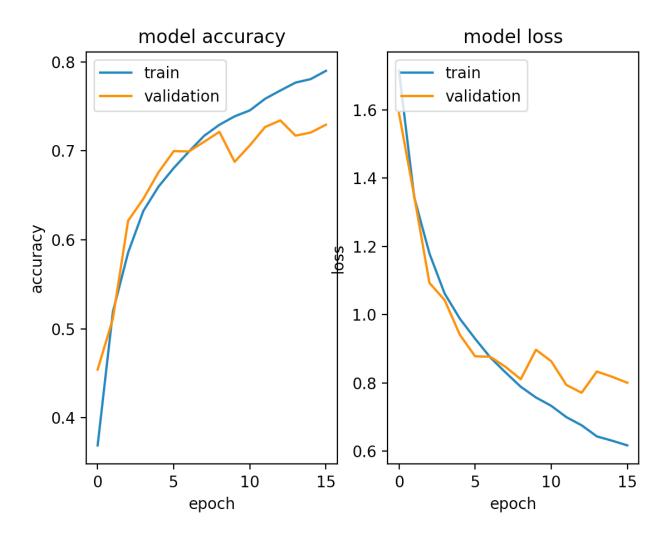


Model: "sequential_8"

Layer (type)	Output Shape	Param #	
conv2d_1 (Conv2D)	(None, 32, 32,	32) 896	
batch_normalization_	1 (Batch (None, 32, 3	2, 32) 128	_
conv2d_2 (Conv2D)	(None, 30, 30,	64) 18496	_
max_pooling2d_1 (M	IaxPooling2 (None, 15	5, 15, 64) 0	_
dropout_1 (Dropout)	(None, 15, 15, 6	4) 0	_
conv2d_3 (Conv2D)	(None, 13, 13,	128) 73856	
max_pooling2d_2 (M	IaxPooling2 (None, 6,	6, 128) 0	
dropout_2 (Dropout)	(None, 6, 6, 128) 0	
flatten_1 (Flatten)	(None, 4608)	0	_
dense_1 (Dense)	(None, 200)	921800	_
dropout_3 (Dropout)	(None, 200)	0	_
dense_2 (Dense)	(None, 100)	20100	_
dropout_4 (Dropout)	(None, 100)	0	_
dense_3 (Dense)	(None, 10)	1010	
Total params: 1,036,2 Trainable params: 1,0 Non-trainable params)36,222		
Epoch 1/20 30149/30149 [====== 0.3689 - val_loss: 1.5 Epoch 2/20 30149/30149 [======	906 - val_accuracy: 0	=======] - 133s 4ms/step - loss .4541 ======] - 109s 4ms/step - loss	
30149/30149 [=====	930 - val_accuracy: 0		: 1.1793 - accuracy:

```
Epoch 4/20
30149/30149 [==
                                              =] - 108s 4ms/step - loss: 1.0616 - accuracy:
0.6323 - val loss: 1.0425 - val accuracy: 0.6459
Epoch 5/20
30149/30149 [============
                                              =] - 108s 4ms/step - loss: 0.9879 - accuracy:
0.6598 - val loss: 0.9408 - val accuracy: 0.6758
Epoch 6/20
= ] - 113s 4ms/step - loss: 0.9294 - accuracy:
0.6806 - val loss: 0.8778 - val accuracy: 0.6996
Epoch 7/20
30149/30149 [======
                                              =] - 106s 4ms/step - loss: 0.8744 - accuracy:
0.6991 - val loss: 0.8763 - val accuracy: 0.6992
Epoch 8/20
==] - 107s 4ms/step - loss: 0.8305 - accuracy:
0.7168 - val loss: 0.8467 - val accuracy: 0.7102
Epoch 9/20
30149/30149 [=======
                                              =] - 110s 4ms/step - loss: 0.7885 - accuracy:
0.7292 - val loss: 0.8107 - val accuracy: 0.7213
Epoch 10/20
30149/30149 [======
                                              =] - 109s 4ms/step - loss: 0.7570 - accuracy:
0.7387 - val loss: 0.8970 - val accuracy: 0.6876
Epoch 11/20
30149/30149 [===========
                                              =] - 116s 4ms/step - loss: 0.7326 - accuracy:
0.7454 - val loss: 0.8633 - val accuracy: 0.7061
Epoch 12/20
= ] - 112s 4ms/step - loss: 0.6994 - accuracy:
0.7584 - val loss: 0.7937 - val accuracy: 0.7267
Epoch 13/20
30149/30149 [======
                                             = ] - 107s 4ms/step - loss: 0.6755 - accuracy:
0.7677 - val loss: 0.7709 - val accuracy: 0.7341
Epoch 14/20
30149/30149 [========
                                             ==] - 108s 4ms/step - loss: 0.6429 - accuracy:
0.7766 - val loss: 0.8332 - val accuracy: 0.7170
Epoch 15/20
== ] - 113s 4ms/step - loss: 0.6303 - accuracy:
0.7806 - val loss: 0.8178 - val accuracy: 0.7205
Epoch 16/20
30149/30149 [======
                                              =] - 124s 4ms/step - loss: 0.6166 - accuracy:
0.7898 - val loss: 0.8003 - val accuracy: 0.7293
=] - 5s 1ms/step
```

- o Train result: train loss: 0.6166 train accuracy: 78.9844
- O Validation result: val loss: 0.8003 val accuracy: 72.9311
- o Test result: test loss: 0.7737 test accuracy: 73.9600



Iteration	Parameters	Accuracy
1	Input layer = 1 Conv	Train = 65.76%
1	Input layer neurons = 32	Validation = 60.72%
	Hidden layers = 1 Conv	Test = 62.18%
	Hidden layers neurons = 16	1650 02.1075
	Hidden layer dropout = 0.5	
	Activation Function = relu	
	Batch size = 32	
	Epochs = 10, early stopped at 9	
2	Input layer = 1 Conv	Train = 52.51%
2	Input layer neurons = 32	Validation = 55.91%
	Hidden layers = 2 Conv	Test = 56.80%
	Hidden layers neurons = [16, 8]	1030 30.0070
	Hidden layers dropout = $[0.5, 0.5]$	
	Activation Function = relu	
	Batch size = 32	
	Epochs = 10	
3	Input layer = 1 Conv	Interrupted training in between
5	Input layer neurons = 32	as accuracy was coming very
	Hidden layers = 2 Conv	low in range of 25-35%
	Hidden layers neurons = [32, 8]	tow in range of 25 5570
	Hidden layers dropout = $[0.5, 0.5]$	
	Activation Function = relu	
	Batch size = 50	
	Epochs = 20, <i>interrupted at 4</i>	
4	Input layer = 1 Conv	Train = 72.52%
7	Input layer neurons = 64	Validation = 65.05%
	Hidden layers = 1 Conv	Test = 67.72%
	Hidden layers neurons = 32	1650 07.7270
	Hidden layers dropout = 0.5	
	Activation Function = relu	
	Batch size = 50	
	Epochs = 20, early stopped at 11	
5	Input layer = 1 Conv	Train = 73.26%
	Input layer neurons = 16	Validation = 65.67%
	Hidden layers = 2 Conv	Test = 67.24%
	Hidden layers neurons = [32, 16]	1000 07.2170
	Hidden layers dropout = $[0.5, 0.5]$	
	Activation Function = relu	
	Batch size = 50	
	Epochs = 20, early stopped at 14	
6	Input layer = 1 Conv	Train = 60.67%
	Input layer neurons = 16	Validation = 67.09%
	Hidden layers = 2 Conv, 1 Dense	Test = 68.84%
	Hidden layers neurons = [32, 16], 100	1651 - 00.0470
	Hidden layers dropout = [0.5, 0.5], 0.4	
	Activation Function = relu	
	Batch size = 50	
	Epochs = 20	
	Epochs = 20	

7	Input layer = 1 Conv	Train = 72.53%
	Input layer neurons = 16	Validation = 71.35%
	Hidden layers = 2 Conv, 2 Dense	Test = 72.68%
	Hidden layers neurons = [32, 64], [100, 100]	
	Hidden layers dropout = $[0.25, 0.25], [0.4, 0.3]$	
	Activation Function = relu	
	Batch size $= 50$	
	Epochs = 20, early stopped at 18	
8	Input layer = 1 Conv	Train = 78.98%
	Input layer neurons = 32	Validation = 72.93%
	Hidden layers = 2 Conv, 2 Dense	Test = 73.96%
	Hidden layers neurons = [64, 128], [200, 100]	
	Hidden layers dropout = $[0.25, 0.25], [0.4, 0.3]$	
	Activation Function = relu	
	Batch size = 50	
	Epochs = 20, early stopped at 17	

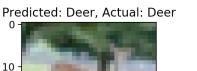
Prediction using best model

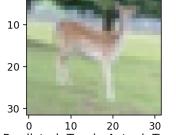
Predictions are made using the best model. The training accuracy of the best model is ~79%.

4597: Predicted: Deer, Actual: Deer

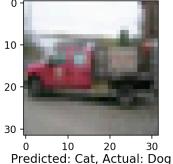
4674: Predicted: Automobile, Actual: Automobile

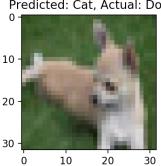
2174: Predicted: Truck, Actual: Truck 4978: Predicted: Horse, Actual: Horse 4172: Predicted: Cat, Actual: Dog 3501: Predicted: Airplane, Actual: Bird



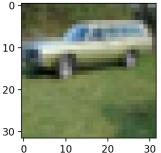


Predicted: Truck, Actual: Truck

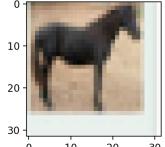




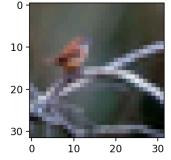
Predicted: Automobile, Actual: Automobile



Predicted: Horse, Actual: Horse



10 20 Predicted: Airplane, Actual: Bird



The result is good but could be better with an improved model.

More models could be trained to obtain a better classifier, but infrastructure limitations (e.g. lack of a powerful computer) has limited the CNN model training scope.

Summary and Conclusion

CNN performed slightly better than the normal NN models created using Keras in Python. The NN model built using H2O in R underperformed relative to the one built using Keras in Python.

- Python Keras NN Model accuracy: Training ~75%, Test ~62%
- Python Keras CNN Model accuracy: Training ~79%, Test ~74%

Using H2O in R, the Model training was limited due to infrastructure limitations:

- o Rectified Linear Units as the activation function
- o Neuron architecture of 3072-128-64-10
- o 20 epochs

Using Keras in Python, further training of more deep models could not be done due to infrastructure limitations.

A major advantage of CNN over NN is that we do not have to flatten the input images to 1D as they are capable of working with image data in 2D. This helps in retaining the "spatial" properties of images.

That is the reason, the data that was available as vectors, could be used in NN, but for CNN, a 2D image data could be used.

However, the dataset used for this project was available as 1D vectors.

The model that performed the best had the following architecture:

- Input layer Conv2D that takes in image data of size (32,32,3) i.e. 32x32 RGB image data
- Batch normalization layer following the input layer to ensure there is not much covariance shift from its output.
- 2 Conv2D hidden layers with 64 and 128 neurons and 'relu' activation function
- Max-pooling was used in the hidden Conv layers to reduce row and column dimensions.
- Dropout was used to inactivate randomly some of the nodes in Conv layers.
- Flatten layer to flatten the 2D data into 1D Vector for the following dense layers.
- 2 Dense layers with 200 and 100 neurons and 'relu' activation function.
- Dropout was used to inactivate randomly some of the nodes in Dense layers.
- Dense output layer with 10 neurons and 'softmax' activation function that is used to obtain probabilities of each of the 10 class.

The models were trained and tuned to find the best classifier

- Validation was done during training with 33% of the training data
- Batch size of 50 was used during training
- Number of epochs were set to 20
- Early stopping was done to stop training if there is no significant improvement in validation accuracy after every 3 epochs. The training stopped after 17 epochs, as a result.
- Checkpointing has been performed to save the best performing model for later use during prediction.

• More improved model could have been found but further training of more deep models could not be done due to infrastructure limitations.

CNN is found to be very useful for the image classification of its high accuracy.

The CNN follows a hierarchical (sequential) model which works on building a network, giving a fully connected layer where all the neurons are connected to each other and the output is processed.

On the input image (32×32) , a filter (say 3x3) can be used to run along all the pixels (rows, columns) of the image which captures the data and then is passed on to the pooling layer where it performs a mathematical computation and gives out a specific result. Here, the filter actually is run across all the values in the pixel matrix and a dot product of the weights and the pixels is calculated.

In data in the convolution layers, is passed into the Rectified Linear Unit (ReLU) activation function to produce corresponding output, which in turn is fed into the next Convolution layer.

Before the output of the hidden convolution layer is fed to the Dense layers, it can be flattened.

Thus, this full 'convoluted' inter-connected network helps in training the model efficiently, making CNN as the best choice for Image Classification.

References

- Image plotting: https://matplotlib.org/api/ as gen/matplotlib.pyplot.imshow.html
- Keras: https://www.tensorflow.org/guide/keras/sequential model