CS/DS541 Deep Learning

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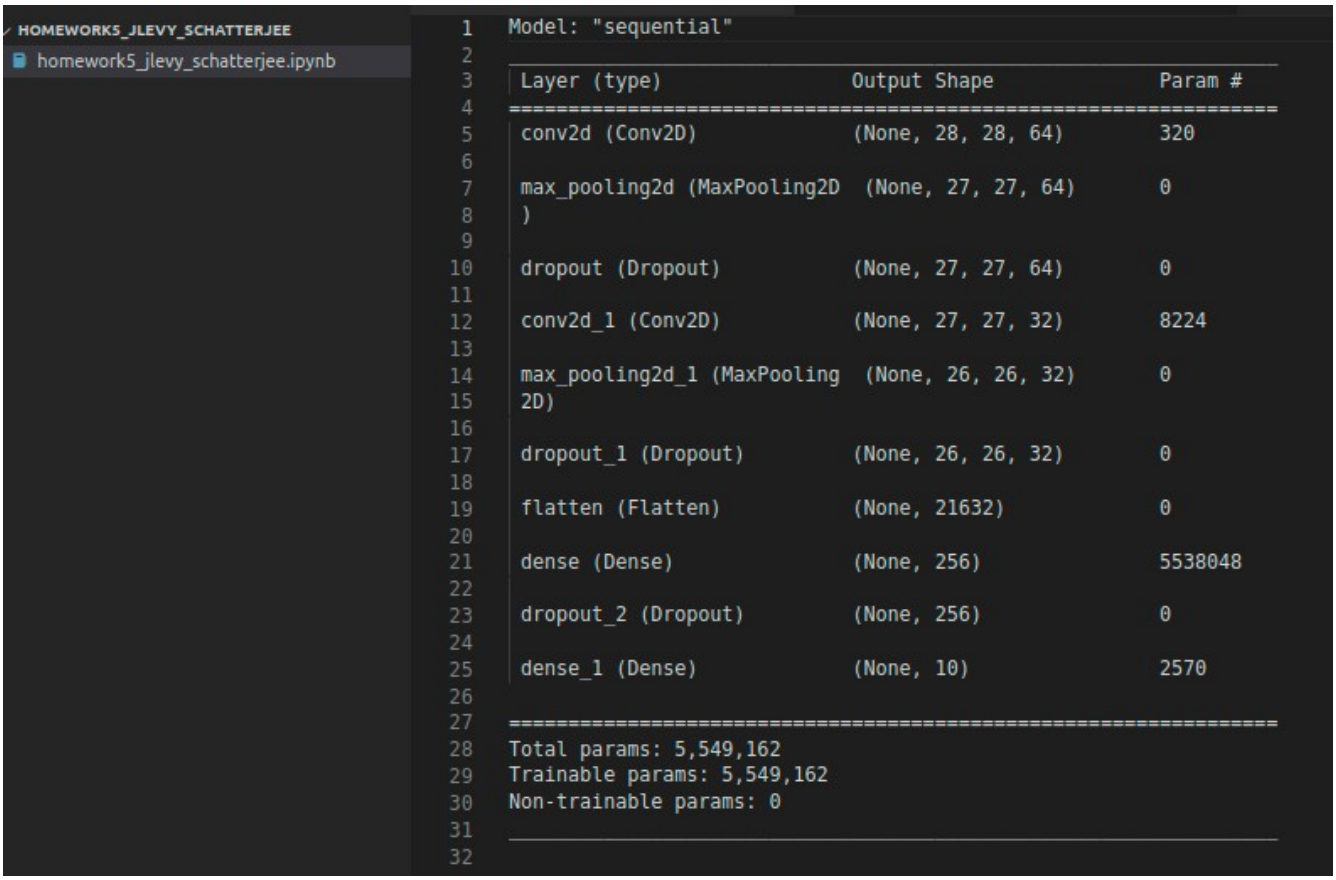
Part1:

Trained for 10 Epochs, and batch size of 64

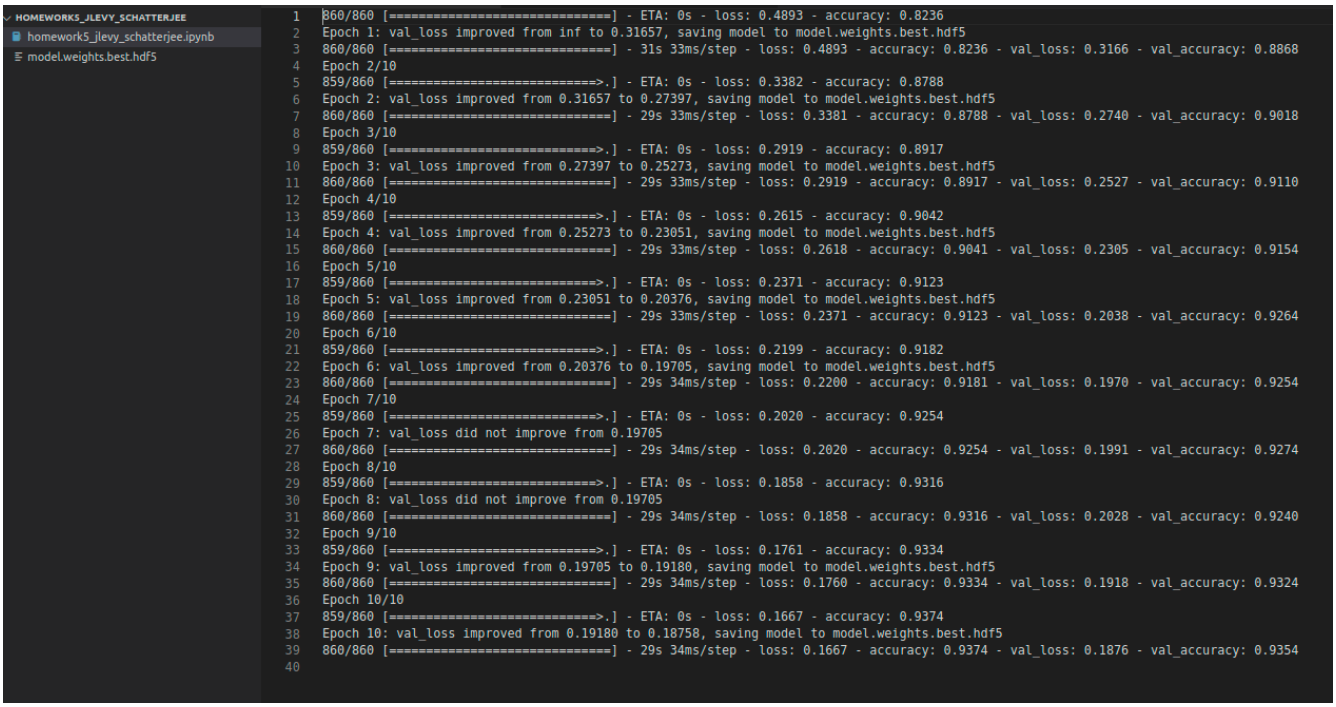


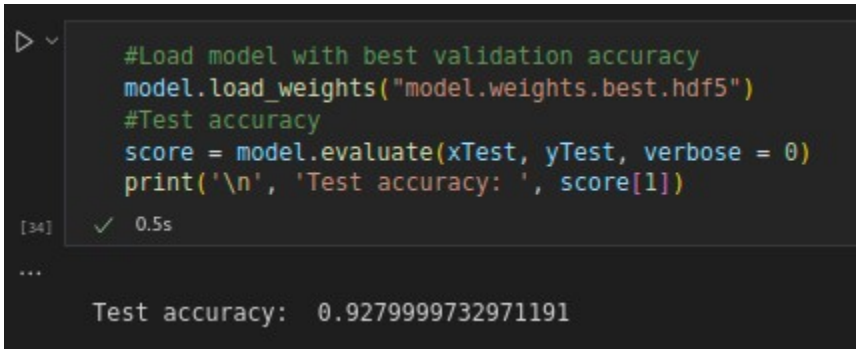
Architecture used:

Convolution layer with 64 filters, each 2x2, stride of 1, without padding activation relu
Max pool with a pooling width of 2x2
Dropout of 0.3, Dropout layer randomly sets input units to 0 with a frequency of rate at each step during training time, which helps prevent overfitting
Convolution layer with 32 filters, each 2x2, stride of 1, without padding activation relu
Max pool with a pooling width of 2x2
Dropout of 0.3,
Flattened all feature maps into one long vector.
Fully-connected layer to map into a 256-dimensional vector
Relu
Dropout of 0.5
Fully-connected layer to map into a 10-dimensional vector.
Softmax.

Results:

Loss and accuracy values over the epochs





Test Accuracy Obtained: 92.79%

Part2:

Trained for 5 Epochs, and batch size of 64

Architecture used:

Convolution layer with 64 filters, each 3x3, stride of 1 no padding.
Max pool with a pooling width of 2x2, stride of 2, no padding.
ReLU.
Flatten the 64 feature maps into one long vector.
Fully-connected layer to map into a 1024-dimensional vector.
ReLU.
Fully-connected layer to map into a 10-dimensional vector.
Softmax.

```
Model: "sequential_3"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 conv2d_6 (Conv2D)           (None, 26, 26, 64)        640

 max_pooling2d_6 (MaxPooling  (None, 13, 13, 64)       0
 2D)

 re_lu (ReLU)                (None, 13, 13, 64)        0

 flatten_3 (Flatten)         (None, 10816)             0

 dense_6 (Dense)             (None, 1024)              11076608

 re_lu_1 (ReLU)              (None, 1024)              0

 dense_7 (Dense)             (None, 10)                10250

 softmax (Softmax)           (None, 10)                0

=================================================================
Total params: 11,087,498
Trainable params: 11,087,498
Non-trainable params: 0
_________________________________________________________________
```

Results:

Loss and accuracy values over the epochs

```
#Compile model
model_p2.compile(loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])

from keras.callbacks import ModelCheckpoint
checkpointer = ModelCheckpoint(filepath = 'model_p2.weights.best.hd5', verbose = 1, save_best_only = True)
#Train the model
model_p2.fit(xTrain, yTrain, batch_size = 64, epochs = 5, validation_data = (xValid, yValid), callbacks = [checkpointer])
```

[48] ✓ 12.7s

```
Epoch 1/5
852/860 [============================>.] - ETA: 0s - loss: 0.0353 - accuracy: 0.9875
Epoch 1: val_loss improved from inf to 0.41162, saving model to model_p2.weights.best.hd5

WARNING:absl:Found untraced functions such as _jit_compiled_convolution_op while saving (showing 1 of 1). These functions will not be directly callable after loading.

INFO:tensorflow:Assets written to: model_p2.weights.best.hd5/assets

INFO:tensorflow:Assets written to: model_p2.weights.best.hd5/assets

860/860 [==============================] - 3s 4ms/step - loss: 0.0351 - accuracy: 0.9875 - val_loss: 0.4116 - val_accuracy: 0.9192
Epoch 2/5
859/860 [============================>.] - ETA: 0s - loss: 0.0221 - accuracy: 0.9932
Epoch 2: val_loss did not improve from 0.41162
860/860 [==============================] - 2s 3ms/step - loss: 0.0221 - accuracy: 0.9932 - val_loss: 0.4152 - val_accuracy: 0.9246
Epoch 3/5
857/860 [============================>.] - ETA: 0s - loss: 0.0158 - accuracy: 0.9949
Epoch 3: val_loss did not improve from 0.41162
860/860 [==============================] - 2s 3ms/step - loss: 0.0159 - accuracy: 0.9949 - val_loss: 0.4592 - val_accuracy: 0.9222
Epoch 4/5
857/860 [============================>.] - ETA: 0s - loss: 0.0205 - accuracy: 0.9929
Epoch 4: val_loss did not improve from 0.41162
860/860 [==============================] - 2s 3ms/step - loss: 0.0206 - accuracy: 0.9929 - val_loss: 0.4818 - val_accuracy: 0.9168
Epoch 5/5
859/860 [============================>.] - ETA: 0s - loss: 0.0111 - accuracy: 0.9963
Epoch 5: val_loss did not improve from 0.41162
860/860 [==============================] - 2s 3ms/step - loss: 0.0111 - accuracy: 0.9963 - val_loss: 0.5166 - val_accuracy: 0.9148

<keras.callbacks.History at 0x7f33cb8d7b50>
```

```
#Test accuracy
score = model_p2.evaluate(xTest, yTest, verbose = 0)
print('\n', 'Test accuracy: ', score[1])
```

[50] ✓ 0.3s

```
Test accuracy:  0.9161999821662903
```

Test Accuracy Obtained: 91.62%

```
This is from Tensorflow Model [3.3310093e-12 1.9235139e-19 5.0639418e-14 1.6371341e-14 7.1424300e-15
 1.6809189e-11 1.6845696e-13 1.0000000e+00 1.4049499e-12 9.4282771e-15]
This is from Numpy Model [3.33100924e-12 1.92351659e-19 5.06393396e-14 1.63713661e-14
 7.14242818e-15 1.68091697e-11 1.68456940e-13 1.00000000e+00
 1.40494739e-12 9.42827208e-15]
This value shows similarity between two layers 100.0
```

Comparison of results from Tensorflow and Numpy code:

We verify the accuracy by predicting a random test data for both the Tensorflow and Numpy model.

Tensorflow:

```
# predicting 1 random sample test data to compare later with numpy forward propagation
yhat1 = model_p2.predict(xTest[998:999,:,:,:])[0]
print(yhat1)
```

[53] ✓ 0.6s

```
1/1 [==============================] - 0s 15ms/step
[3.3310093e-12 1.9235139e-19 5.0639418e-14 1.6371341e-14 7.1424300e-15
 1.6809189e-11 1.6845696e-13 1.0000000e+00 1.4049499e-12 9.4282771e-15]
```

Numpy:

```
# Working on the same test data xTest[998]

yhat = softmax(h3)
#print(yhat.Shape)
yhat2 = yhat
yhat2 = yhat.reshape(-1)
print(yhat2)
```

[64] ✓ 0.1s

```
[3.33100924e-12 1.92351659e-19 5.06393396e-14 1.63713661e-14
 7.14242818e-15 1.68091697e-11 1.68456940e-13 1.00000000e+00
 1.40494739e-12 9.42827208e-15]
```

```
This is from Tensorflow [3.3310093e-12 1.9235139e-19 5.0639418e-14 1.6371341e-14 7.1424300e-15
 1.6809189e-11 1.6845696e-13 1.0000000e+00 1.4049499e-12 9.4282771e-15]
This is from Network through Numpy [3.33100924e-12 1.92351659e-19 5.06393396e-14 1.63713661e-14
 7.14242818e-15 1.68091697e-11 1.68456940e-13 1.00000000e+00
 1.40494739e-12 9.42827208e-15]
This is accuracy which shows how two layers are similar 100.0
```

Results:

Both outputs have same values leading to a 100% similarity in results

The following plot shows the same.