## cnn-dl

## April 17, 2023

```
[18]: import tensorflow as tf
     from tensorflow import keras
     from tensorflow.keras import layers
     from tensorflow.keras.datasets import cifar10
     from tensorflow.keras.utils import to_categorical
     from tensorflow.keras.callbacks import EarlyStopping
[3]: (x_train, y_train), (x_test, y_test) = cifar10.load_data()
     Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz
     [4]: x_train = x_train.astype("float32") / 255.0
     x_test = x_test.astype("float32") / 255.0
[5]: y_train = to_categorical(y_train)
     y_test = to_categorical(y_test)
[6]: model = keras.Sequential(
             layers.Conv2D(32, kernel_size=(3, 3), activation="relu", u
      →input_shape=(32, 32, 3)),
             layers.MaxPooling2D(pool_size=(2, 2)),
             layers.Conv2D(64, kernel_size=(3, 3), activation="relu"),
             layers.MaxPooling2D(pool_size=(2, 2)),
             layers.Flatten(),
             layers.Dense(10, activation="softmax"),
         ]
     )
```

## 0.0.1 1)Method

```
[]: sgd = keras.optimizers.SGD(lr=0.01)
model.compile(loss="categorical_crossentropy", optimizer=sgd,
→metrics=["accuracy"])
```

WARNING:absl:`lr` is deprecated in Keras optimizer, please use `learning\_rate` or use the legacy optimizer, e.g.,tf.keras.optimizers.legacy.SGD.

```
Epoch 1/15
accuracy: 0.8155 - val_loss: 1.0158 - val_accuracy: 0.7016
782/782 [========== ] - 71s 91ms/step - loss: 0.5105 -
accuracy: 0.8251 - val_loss: 1.4288 - val_accuracy: 0.6284
782/782 [============= ] - 72s 92ms/step - loss: 0.5066 -
accuracy: 0.8282 - val_loss: 1.1546 - val_accuracy: 0.6719
Epoch 4/15
782/782 [============= ] - 71s 91ms/step - loss: 0.4943 -
accuracy: 0.8334 - val loss: 1.0403 - val accuracy: 0.7066
Epoch 5/15
782/782 [============ ] - 70s 90ms/step - loss: 0.4837 -
accuracy: 0.8348 - val_loss: 1.0528 - val_accuracy: 0.7006
Epoch 6/15
782/782 [============ ] - 71s 91ms/step - loss: 0.4799 -
accuracy: 0.8369 - val_loss: 0.9942 - val_accuracy: 0.7152
Epoch 7/15
782/782 [============ ] - 71s 91ms/step - loss: 0.4718 -
accuracy: 0.8390 - val_loss: 1.0083 - val_accuracy: 0.7145
Epoch 8/15
accuracy: 0.8400 - val_loss: 1.2644 - val_accuracy: 0.6668
Epoch 9/15
782/782 [=========== ] - 70s 89ms/step - loss: 0.4637 -
accuracy: 0.8429 - val_loss: 1.0594 - val_accuracy: 0.6983
Epoch 10/15
782/782 [============ ] - 71s 91ms/step - loss: 0.4628 -
accuracy: 0.8411 - val_loss: 1.0478 - val_accuracy: 0.7036
Epoch 11/15
782/782 [============= ] - 71s 91ms/step - loss: 0.4562 -
accuracy: 0.8442 - val_loss: 1.0478 - val_accuracy: 0.7085
Epoch 12/15
782/782 [============= ] - 69s 88ms/step - loss: 0.4535 -
accuracy: 0.8442 - val_loss: 1.0413 - val_accuracy: 0.7073
Epoch 13/15
accuracy: 0.8467 - val_loss: 1.0796 - val_accuracy: 0.6994
Epoch 14/15
782/782 [============ ] - 69s 89ms/step - loss: 0.4468 -
accuracy: 0.8460 - val loss: 1.0542 - val accuracy: 0.7022
Epoch 15/15
782/782 [============= ] - 71s 91ms/step - loss: 0.4457 -
```

```
accuracy: 0.8477 - val_loss: 1.2613 - val_accuracy: 0.6679
[]: <keras.callbacks.History at 0x7f32507626a0>
[]: test_loss, test_acc = model.evaluate(x_test, y_test)
   print(f"Test accuracy: {test_acc}")
  accuracy: 0.6679
  Test accuracy: 0.667900025844574
     Finetune the hyperparameters to improve the performance of
     the CNN
  1.0.1 2) Method
[7]: model.compile(loss="categorical_crossentropy", optimizer="adam", ___
    →metrics=["accuracy"])
[8]: model.fit(x_train, y_train, batch_size=64, epochs=10, validation_data=(x_test,__

y test))
  Epoch 1/10
  accuracy: 0.4511 - val_loss: 1.2891 - val_accuracy: 0.5381
  Epoch 2/10
  accuracy: 0.5808 - val_loss: 1.1184 - val_accuracy: 0.6150
  Epoch 3/10
  782/782 [============ ] - 71s 91ms/step - loss: 1.0717 -
  accuracy: 0.6318 - val_loss: 1.0415 - val_accuracy: 0.6450
  Epoch 4/10
  accuracy: 0.6590 - val_loss: 1.0068 - val_accuracy: 0.6544
  accuracy: 0.6802 - val_loss: 0.9966 - val_accuracy: 0.6601
  Epoch 6/10
  accuracy: 0.6948 - val_loss: 0.9450 - val_accuracy: 0.6745
  accuracy: 0.7116 - val_loss: 0.9646 - val_accuracy: 0.6751
  Epoch 8/10
  782/782 [============ ] - 71s 91ms/step - loss: 0.8150 -
  accuracy: 0.7216 - val_loss: 0.9199 - val_accuracy: 0.6901
```

Epoch 9/10

```
782/782 [============== ] - 69s 89ms/step - loss: 0.7828 -
   accuracy: 0.7318 - val_loss: 0.9020 - val_accuracy: 0.6968
   Epoch 10/10
   782/782 [============ ] - 71s 91ms/step - loss: 0.7576 -
   accuracy: 0.7376 - val_loss: 0.8907 - val_accuracy: 0.6960
[8]: <keras.callbacks.History at 0x7f32508f4700>
[9]: test_loss, test_acc = model.evaluate(x_test, y_test)
    print(f"Test accuracy: {test_acc}")
   accuracy: 0.6960
   Test accuracy: 0.6959999799728394
   1.0.2 3) Method
[]: early_stop = EarlyStopping(monitor='val_loss', patience=3)
[]: model.fit(x_train, y_train, batch_size=64, epochs=100, validation_data=(x_test,__

y_test), callbacks=[early_stop])
   Epoch 1/100
   accuracy: 0.8063 - val_loss: 0.1603 - val_accuracy: 0.7100
   782/782 [========== ] - 71s 91ms/step - loss: 0.1134 -
   accuracy: 0.8078 - val_loss: 0.1619 - val_accuracy: 0.7072
   Epoch 3/100
   782/782 [============= ] - 69s 89ms/step - loss: 0.1124 -
   accuracy: 0.8100 - val_loss: 0.1592 - val_accuracy: 0.7181
   Epoch 4/100
   accuracy: 0.8140 - val_loss: 0.1586 - val_accuracy: 0.7175
   Epoch 5/100
   782/782 [============ ] - 72s 92ms/step - loss: 0.1097 -
   accuracy: 0.8167 - val_loss: 0.1599 - val_accuracy: 0.7152
   Epoch 6/100
   782/782 [============ ] - 72s 92ms/step - loss: 0.1087 -
   accuracy: 0.8194 - val_loss: 0.1624 - val_accuracy: 0.7123
   Epoch 7/100
   782/782 [============= ] - 71s 91ms/step - loss: 0.1076 -
   accuracy: 0.8194 - val_loss: 0.1618 - val_accuracy: 0.7153
[]: <keras.callbacks.History at 0x7f32506070a0>
[]: test_loss, test_acc = model.evaluate(x_test, y_test)
    print(f"Test accuracy: {test_acc}")
```

```
accuracy: 0.7153
  Test accuracy: 0.7153000235557556
  1.0.3 4) Method
[]: model.compile(loss="binary_crossentropy", optimizer="adam", u
   →metrics=["accuracy"])
[]: model.fit(x_train, y_train, batch_size=128, epochs=15, validation_data=(x_test,__
   →y_test))
  Epoch 1/15
  accuracy: 0.7592 - val_loss: 0.1613 - val_accuracy: 0.7025
  Epoch 2/15
  accuracy: 0.7648 - val_loss: 0.1574 - val_accuracy: 0.7089
  Epoch 3/15
  accuracy: 0.7678 - val_loss: 0.1571 - val_accuracy: 0.7064
  Epoch 4/15
  accuracy: 0.7716 - val_loss: 0.1557 - val_accuracy: 0.7092
  Epoch 5/15
  accuracy: 0.7767 - val_loss: 0.1567 - val_accuracy: 0.7080
  Epoch 6/15
  accuracy: 0.7798 - val_loss: 0.1539 - val_accuracy: 0.7128
  Epoch 7/15
  accuracy: 0.7837 - val_loss: 0.1543 - val_accuracy: 0.7148
  Epoch 8/15
  accuracy: 0.7867 - val_loss: 0.1547 - val_accuracy: 0.7168
  Epoch 9/15
  accuracy: 0.7891 - val_loss: 0.1531 - val_accuracy: 0.7183
  Epoch 10/15
  accuracy: 0.7919 - val_loss: 0.1553 - val_accuracy: 0.7126
  Epoch 11/15
  accuracy: 0.7949 - val_loss: 0.1536 - val_accuracy: 0.7195
```

accuracy: 0.7968 - val\_loss: 0.1543 - val\_accuracy: 0.7148

```
[]: test_loss, test_acc = model.evaluate(x_test, y_test)
print(f"Test accuracy: {test_acc}")
```

Test accuracy: 0.7193999886512756

After applying multiple method the last method has highest accuracy so we consider this model compiler and fit for training