NagarajVinay_Assignment_6_2b

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0.1 Assignment 6.2b

0.1.1 CIFAR10 Classification with dropout and augmentation

```
[1]: from keras.datasets import cifar10
     from keras.utils import to_categorical
     from keras.preprocessing.image import ImageDataGenerator
     (x_train, y_train), (x_test, y_test) = cifar10.load_data()
[2]: x_train.shape, y_train.shape
[2]: ((50000, 32, 32, 3), (50000, 1))
[3]: x_test.shape, y_test.shape
[3]: ((10000, 32, 32, 3), (10000, 1))
[4]: # Preprocess the data (these are NumPy arrays)
     x_train = x_train.astype("float32")
     x_test = x_test.astype("float32")
     y_train = to_categorical(y_train)
     y_test = to_categorical(y_test)
[5]: # Reserve 10,000 samples for validation
     x_val = x_train[-10000:]
     y_val = y_train[-10000:]
     x_train_2 = x_train[:-10000]
     y_train_2 = y_train[:-10000]
[6]: train_datagen = ImageDataGenerator(rescale=1./255,
                                       rotation_range=40,
                                       width_shift_range=0.2,
                                       height_shift_range=0.2,
                                       shear_range=0.2,
                                       zoom_range=0.2,
                                       horizontal_flip=True)
```

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test_datagen = ImageDataGenerator(rescale=1./255)
train_generator = train_datagen.flow(x_train_2, y_train_2, batch_size=32)
validation_generator = train_datagen.flow(x_val, y_val, batch_size=32)
```

```
[7]: #instantiate the model
from keras import models
from keras import layers

model = models.Sequential()
model.add(layers.Conv2D(32, (3,3), activation='relu', input_shape=(32,32,3)))
model.add(layers.MaxPooling2D(2,2))
model.add(layers.Conv2D(64, (3,3), activation='relu'))
model.add(layers.MaxPooling2D(2,2))
model.add(layers.Conv2D(64, (3,3), activation='relu'))
model.add(layers.MaxPooling2D(2,2))
model.add(layers.Flatten())
model.add(layers.Dropout(0.5))
model.add(layers.Dense(64, activation='relu'))
model.add(layers.Dense(10, activation='relu'))
model.add(layers.Dense(10, activation='softmax'))
model.summary()
```

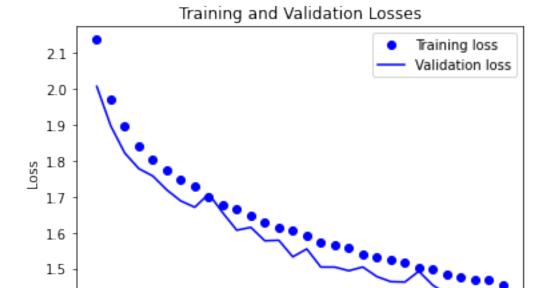
Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 30, 30, 32)	896
max_pooling2d (MaxPooling2D)	(None, 15, 15, 32)	0
conv2d_1 (Conv2D)	(None, 13, 13, 64)	18496
max_pooling2d_1 (MaxPooling2	(None, 6, 6, 64)	0
conv2d_2 (Conv2D)	(None, 4, 4, 64)	36928
max_pooling2d_2 (MaxPooling2	(None, 2, 2, 64)	0
flatten (Flatten)	(None, 256)	0
dropout (Dropout)	(None, 256)	0
dense (Dense)	(None, 64)	16448

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dense_1 (Dense)
                         (None, 10)
                                            650
   ______
   Total params: 73,418
   Trainable params: 73,418
   Non-trainable params: 0
[8]: from keras import optimizers
   model.compile(optimizer=optimizers.RMSprop(lr=1e-4),
             loss='categorical_crossentropy',
             metrics=['accuracy'])
[9]: history = model.fit_generator(train_generator,
                         steps_per_epoch=len(x_train_2) / 32,
                         epochs=30,
                         validation_data=validation_generator,
                         validation_steps=len(x_val) / 32)
   /opt/conda/lib/python3.8/site-
   packages/tensorflow/python/keras/engine/training.py:1844: UserWarning:
   `Model.fit_generator` is deprecated and will be removed in a future version.
   Please use `Model.fit`, which supports generators.
    warnings.warn('`Model.fit_generator` is deprecated and '
   Epoch 1/30
   accuracy: 0.1609 - val_loss: 2.0055 - val_accuracy: 0.2568
   Epoch 2/30
   accuracy: 0.2504 - val_loss: 1.8961 - val_accuracy: 0.2979
   Epoch 3/30
   accuracy: 0.2817 - val_loss: 1.8206 - val_accuracy: 0.3249
   Epoch 4/30
   accuracy: 0.3087 - val_loss: 1.7776 - val_accuracy: 0.3418
   Epoch 5/30
   accuracy: 0.3301 - val_loss: 1.7570 - val_accuracy: 0.3641
   Epoch 6/30
   1250/1250 [============= ] - 49s 39ms/step - loss: 1.7741 -
   accuracy: 0.3423 - val_loss: 1.7185 - val_accuracy: 0.3729
   Epoch 7/30
   1250/1250 [============= ] - 51s 41ms/step - loss: 1.7549 -
   accuracy: 0.3523 - val_loss: 1.6878 - val_accuracy: 0.3850
   Epoch 8/30
   1250/1250 [============== ] - 43s 35ms/step - loss: 1.7325 -
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accuracy: 0.3604 - val_loss: 1.6704 - val_accuracy: 0.3969
Epoch 9/30
accuracy: 0.3740 - val_loss: 1.7073 - val_accuracy: 0.3863
Epoch 10/30
accuracy: 0.3836 - val_loss: 1.6548 - val_accuracy: 0.4063
Epoch 11/30
accuracy: 0.3881 - val_loss: 1.6066 - val_accuracy: 0.4160
Epoch 12/30
accuracy: 0.3975 - val_loss: 1.6146 - val_accuracy: 0.4087
Epoch 13/30
accuracy: 0.4130 - val_loss: 1.5774 - val_accuracy: 0.4256
Epoch 14/30
accuracy: 0.4133 - val_loss: 1.5786 - val_accuracy: 0.4354
Epoch 15/30
accuracy: 0.4164 - val_loss: 1.5328 - val_accuracy: 0.4449
Epoch 16/30
accuracy: 0.4133 - val_loss: 1.5546 - val_accuracy: 0.4447
Epoch 17/30
accuracy: 0.4322 - val_loss: 1.5044 - val_accuracy: 0.4636
accuracy: 0.4311 - val_loss: 1.5043 - val_accuracy: 0.4653
Epoch 19/30
accuracy: 0.4377 - val_loss: 1.4941 - val_accuracy: 0.4649
Epoch 20/30
accuracy: 0.4423 - val loss: 1.5046 - val accuracy: 0.4630
Epoch 21/30
accuracy: 0.4486 - val_loss: 1.4780 - val_accuracy: 0.4738
Epoch 22/30
accuracy: 0.4561 - val_loss: 1.4639 - val_accuracy: 0.4764
Epoch 23/30
1250/1250 [============= ] - 43s 34ms/step - loss: 1.5201 -
accuracy: 0.4523 - val_loss: 1.4628 - val_accuracy: 0.4729
Epoch 24/30
1250/1250 [============= ] - 43s 35ms/step - loss: 1.4954 -
```

```
accuracy: 0.4634 - val_loss: 1.4927 - val_accuracy: 0.4640
    Epoch 25/30
    1250/1250 [============= ] - 43s 34ms/step - loss: 1.4944 -
    accuracy: 0.4624 - val_loss: 1.4540 - val_accuracy: 0.4787
    Epoch 26/30
    accuracy: 0.4685 - val_loss: 1.4296 - val_accuracy: 0.4884
    Epoch 27/30
    1250/1250 [============ ] - 43s 34ms/step - loss: 1.4768 -
    accuracy: 0.4696 - val_loss: 1.4212 - val_accuracy: 0.4955
    Epoch 28/30
    accuracy: 0.4696 - val_loss: 1.4201 - val_accuracy: 0.4955
    Epoch 29/30
    accuracy: 0.4737 - val_loss: 1.4032 - val_accuracy: 0.5011
    Epoch 30/30
    1250/1250 [============= ] - 43s 34ms/step - loss: 1.4503 -
    accuracy: 0.4797 - val_loss: 1.4026 - val_accuracy: 0.4942
[10]: import matplotlib.pyplot as plt
    train_loss = history.history['loss']
    val_loss = history.history['val_loss']
    epochs = range(1, len(history.history['loss']) + 1)
    plt.plot(epochs, train_loss, 'bo', label='Training loss')
    plt.plot(epochs, val_loss, 'b', label='Validation loss')
    plt.title('Training and Validation Losses')
    plt.xlabel('Epochs')
    plt.ylabel('Loss')
    plt.legend()
    plt.show()
    plt.savefig('results/6_2b_lossplot.png')
```



Epochs

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```
[11]: import matplotlib.pyplot as plt

train_loss = history.history['accuracy']

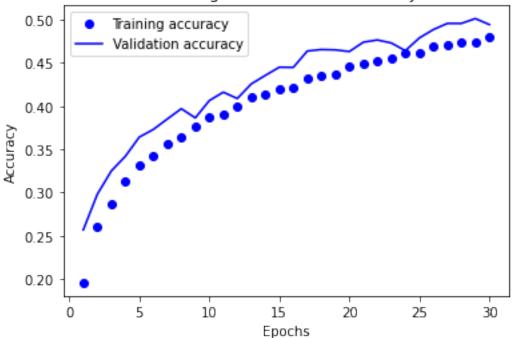
val_loss = history.history['val_accuracy']

epochs = range(1, len(history.history['accuracy']) + 1)

plt.plot(epochs, train_loss, 'bo', label='Training accuracy')
plt.plot(epochs, val_loss, 'b', label='Validation accuracy')
plt.title('Training and Validation Accuracy')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()

plt.show()
plt.savefig('results/6_2b_accplot.png')
```





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```
Epoch 2/16
1562/1562 [============= ] - 44s 28ms/step - loss: 1.4457 -
accuracy: 0.4815
Epoch 3/16
1562/1562 [============== ] - 45s 28ms/step - loss: 1.4408 -
accuracy: 0.4844
Epoch 4/16
accuracy: 0.4864
Epoch 5/16
accuracy: 0.4913
Epoch 6/16
accuracy: 0.4875
Epoch 7/16
1562/1562 [============= ] - 45s 29ms/step - loss: 1.4113 -
accuracy: 0.4946
Epoch 8/16
accuracy: 0.4961
Epoch 9/16
accuracy: 0.5059
Epoch 10/16
accuracy: 0.5019
Epoch 11/16
accuracy: 0.5013
Epoch 12/16
1562/1562 [=============== ] - 57s 36ms/step - loss: 1.3925 -
accuracy: 0.5038
Epoch 13/16
1562/1562 [============= - - 57s 37ms/step - loss: 1.3662 -
accuracy: 0.5132
Epoch 14/16
accuracy: 0.5171
Epoch 15/16
1562/1562 [============= ] - 57s 36ms/step - loss: 1.3738 -
accuracy: 0.5083
Epoch 16/16
1562/1562 [============= - - 57s 36ms/step - loss: 1.3681 -
accuracy: 0.5101
accuracy: 0.3377
```

```
[13]: model.save('results/6_2b_model.h5')
[14]: prediction_results = model.predict(x_test)
[15]: #write metrics to file
    with open('results/6_2b_metrics.txt', 'w') as f:
        f.write('Training Loss: {}'.format(str(history.history['loss'])))
        f.write('\nTraining Accuracy: {}'.format(str(history.history['accuracy'])))
        f.write('\nTest Loss: {}'.format(results[0]))
        f.write('\nTest Accuracy: {}'.format(results[1]))
[16]: #write metrics to file
    with open('results/6_2b_metrics.txt', 'w') as f:
        f.write('Training Loss: {}'.format(str(history.history['loss'])))
        f.write('\nTraining Accuracy: {}'.format(str(history.history['accuracy'])))
        f.write('\nTest Loss: {}'.format(results[0]))
        f.write('\nTest Accuracy: {}'.format(results[1]))
[ ]:
```