

SCREENSHOTS

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]]: #importing the Libraries
import numpy as np
from keras.datasets import cifar10
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.constraints import maxnorm
from keras.optimizers import SGD
from keras.layers.convolutional import Conv2D, MaxPooling2D
from keras.utils import np_utils

#Set random seed value
np.random.seed(7)
#Load and preprocess data
(X_train, y_train), (X_test, y_test) = cifar10.load_data()

X_train = X_train.astype('float32') / 255.0
X_test = X_test.astype('float32') / 255.0

y_train = np_utils.to_categorical(y_train)
y_test = np_utils.to_categorical(y_test)
num_classes = y_test.shape[1]

#Create Model
model = Sequential()
model.add(Conv2D(32, (3, 3), input_shape=(32, 32, 3), padding='same', activation='relu', kernel_constraint=maxnorm(3)))
model.add(Dropout(0.2))
model.add(Conv2D(32, (3, 3), activation='relu', padding='same', kernel_constraint=maxnorm(3)))
model.add(MaxPooling2D(pool_size=(2, 2), padding='same'))
model.add(Flatten())
model.add(Dense(512, activation='relu', kernel_constraint=maxnorm(3)))
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax'))
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#Compile model
sgd = SGD(learning_rate=0.01, momentum=0.9, decay=1e-6)
model.compile(loss='categorical_crossentropy', optimizer=sgd, metrics=['accuracy'])
print(model.summary())

#Train model
epochs = 5
batch_size = 32
model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=epochs, batch_size=batch_size)
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #
conv2d_2 (Conv2D)	(None, 32, 32, 32)	896
dropout_2 (Dropout)	(None, 32, 32, 32)	0
conv2d_3 (Conv2D)	(None, 32, 32, 32)	9248
max_pooling2d_1 (MaxPooling 2D)	(None, 16, 16, 32)	0
flatten_1 (Flatten)	(None, 8192)	0
dense_2 (Dense)	(None, 512)	4194816
dropout_3 (Dropout)	(None, 512)	0
dense_3 (Dense)	(None, 10)	5130

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Total params: 4,210,090
Trainable params: 4,210,090
Non-trainable params: 0

None
Model: "sequential_1"

Layer (type)	Output Shape	Param #
conv2d_2 (Conv2D)	(None, 32, 32, 32)	896
dropout_2 (Dropout)	(None, 32, 32, 32)	0
conv2d_3 (Conv2D)	(None, 32, 32, 32)	9248
max_pooling2d_1 (MaxPooling 2D)	(None, 16, 16, 32)	0
flatten_1 (Flatten)	(None, 8192)	0
dense_2 (Dense)	(None, 512)	4194816
dropout_3 (Dropout)	(None, 512)	0
dense_3 (Dense)	(None, 10)	5130

=====
Total params: 4,210,090
Trainable params: 4,210,090
Non-trainable params: 0

None
Epoch 1/5
1563/1563 [=====] - 164s 104ms/step - loss: 1.7320 - accuracy: 0.3701 - val_loss: 1.4194 - val_accuracy: 0.4944
Epoch 2/5
1563/1563 [=====] - 165s 105ms/step - loss: 1.3552 - accuracy: 0.5128 - val_loss: 1.2045 - val_accuracy: 0.5676

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Epoch 3/5
1563/1563 [=====] - 165s 106ms/step - loss: 1.1688 - accuracy: 0.5842 - val_loss: 1.0648 - val_accuracy: 0.6252
Epoch 4/5
1563/1563 [=====] - 160s 102ms/step - loss: 1.0351 - accuracy: 0.6336 - val_loss: 0.9853 - val_accuracy: 0.6544
Epoch 5/5
1563/1563 [=====] - 165s 106ms/step - loss: 0.9280 - accuracy: 0.6732 - val_loss: 0.9793 - val_accuracy: 0.6610

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Out[5]: <keras.callbacks.History at 0x7f67c770eb80>

In []:

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In [ ]: scores = model.evaluate(X_test, y_test, verbose=0)
print("Accuracy: %.2f%%" % (scores[1]*100))

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Accuracy: 66.10%

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model.add(Dense(128, activation='relu', kernel_constraint=maxnorm(3)))
model.add(Dropout(0.2))
model.add(Dense(512, activation='relu', kernel_constraint=maxnorm(3)))
model.add(Dropout(0.2))
model.add(Dense(num_classes, activation='softmax'))

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[ ]: epochs = 5
learning_rate = 0.01
decay_rate = learning_rate / epochs
sgd = SGD(lr=learning_rate, momentum=0.9, decay=decay_rate, nesterov=False)
model.compile(loss='categorical_crossentropy', optimizer=sgd, metrics=['accuracy'])
print(model.summary())

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Model: "sequential_2"

Layer (type)	Output Shape	Param #
conv2d_4 (Conv2D)	(None, 32, 32, 32)	896
dropout_4 (Dropout)	(None, 32, 32, 32)	0
conv2d_5 (Conv2D)	(None, 32, 32, 32)	9248
max_pooling2d_2 (MaxPooling2D)	(None, 16, 16, 32)	0
conv2d_6 (Conv2D)	(None, 16, 16, 64)	18496
dropout_5 (Dropout)	(None, 16, 16, 64)	0
conv2d_7 (Conv2D)	(None, 16, 16, 64)	36928
max_pooling2d_3 (MaxPooling2D)	(None, 8, 8, 64)	0

conv2d_8 (Conv2D)	(None, 8, 8, 128)	73856
dropout_6 (Dropout)	(None, 8, 8, 128)	0
conv2d_9 (Conv2D)	(None, 8, 8, 128)	147584
max_pooling2d_4 (MaxPooling 2D)	(None, 4, 4, 128)	0
flatten_2 (Flatten)	(None, 2048)	0
dropout_7 (Dropout)	(None, 2048)	0
dense_4 (Dense)	(None, 1024)	2098176
dropout_8 (Dropout)	(None, 1024)	0
dense_5 (Dense)	(None, 512)	524800
dropout_9 (Dropout)	(None, 512)	0
dense_6 (Dense)	(None, 10)	5130

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=====
Total params: 2,915,114
Trainable params: 2,915,114
Non-trainable params: 0

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None

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|: history = model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=epochs, batch_size=32)

scores = model.evaluate(X_test, y_test, verbose=0)
print("Accuracy: %.2f%%" % (scores[1] * 100))

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Epoch 1/5
1563/1563 [=====] - 320s 204ms/step - loss: 1.8690 - accuracy: 0.3094 - val_loss: 1.5686 - val_accuracy: 0.4288
Epoch 2/5
1563/1563 [=====] - 319s 204ms/step - loss: 1.4913 - accuracy: 0.4537 - val_loss: 1.3887 - val_accuracy: 0.4972
Epoch 3/5
1563/1563 [=====] - 319s 204ms/step - loss: 1.3537 - accuracy: 0.5097 - val_loss: 1.3130 - val_accuracy: 0.5285
Epoch 4/5
1563/1563 [=====] - 318s 203ms/step - loss: 1.2687 - accuracy: 0.5440 - val_loss: 1.1876 - val_accuracy: 0.5719
Epoch 5/5
1563/1563 [=====] - 318s 204ms/step - loss: 1.2044 - accuracy: 0.5676 - val_loss: 1.1627 - val_accuracy: 0.5805
Accuracy: 58.05%

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```
import numpy as np
predictions = model.predict(X_test[:4])
predicted_labels = np.argmax(predictions, axis=1)
actual_labels = np.argmax(y_test[:4], axis=1)

print("Predicted labels:", predicted_labels)
print("Actual labels:  ", actual_labels)
```

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1/1 [=====] - 0s 25ms/step
Predicted labels: [3 1 8 0]
Actual labels:   [3 8 8 0]
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