NEURAL NETWORKS ASSIGNMENT-3

Deep Learning Image Classification with CNN

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GITHUB LINK: https://github.com/sxm26790/ASSIGNMENT3_NEURAL

To perform image classification with a CNN, we'll use the Keras library, which provides an easy-to-use API for building and training deep learning models. Let's go through the implementation step by step:

Import the required 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
```

```
np.random.seed(7)
```

Load and preprocess the data: We'll use the CIFAR-10 dataset, which contains 50,000 training images and 10,000 test images of 10 different classes.

```
(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
    170498071/170498071 [==============================] - 3s 0us/step
```

```
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]
```

Build the CNN 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'))
```

```
sgd = SGD(learning_rate=0.01, momentum=0.9, decay=1e-6)
model.compile(loss='categorical_crossentropy', optimizer=sgd, metrics=['accuracy'])
print(model.summary())
```

```
    Model: "sequential"
    _________________________________________________________________
     Layer (type)                Output Shape              Param #
    =================================================================
     conv2d (Conv2D)             (None, 32, 32, 32)        896

     dropout (Dropout)           (None, 32, 32, 32)        0

     conv2d_1 (Conv2D)           (None, 32, 32, 32)        9248

     max_pooling2d (MaxPooling2D  (None, 16, 16, 32)       0
     )

     flatten (Flatten)           (None, 8192)              0

     dense (Dense)               (None, 512)               4194816

     dropout_1 (Dropout)         (None, 512)               0

     dense_1 (Dense)             (None, 10)                5130
```

```
=================================================================
Total params: 4,210,090
Trainable params: 4,210,090
Non-trainable params: 0
_________________________________________________________________
None
```

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

```
Epoch 1/5
1563/1563 [==============================] - 164s 105ms/step - loss: 1.7276 - accuracy: 0.3737 - val_loss: 1.4764 - val_accuracy: 0.4738
Epoch 2/5
1563/1563 [==============================] - 168s 108ms/step - loss: 1.4083 - accuracy: 0.4939 - val_loss: 1.2520 - val_accuracy: 0.5621
Epoch 3/5
1563/1563 [==============================] - 168s 108ms/step - loss: 1.2226 - accuracy: 0.5638 - val_loss: 1.1403 - val_accuracy: 0.5962
Epoch 4/5
1563/1563 [==============================] - 167s 107ms/step - loss: 1.0943 - accuracy: 0.6109 - val_loss: 1.0451 - val_accuracy: 0.6321
Epoch 5/5
1563/1563 [==============================] - 169s 108ms/step - loss: 0.9879 - accuracy: 0.6515 - val_loss: 1.1353 - val_accuracy: 0.6127
<keras.callbacks.History at 0x781917f1de40>
```

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

```
Accuracy: 61.27%
```

```
import numpy as np
from keras.datasets import cifar10
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.layers.convolutional import Conv2D, MaxPooling2D
from keras.constraints import maxnorm
from keras.utils import np_utils
from keras.optimizers import SGD

# Fix random seed for reproducibility
np.random.seed(7)

# Load data
(X_train, y_train), (X_test, y_test) = cifar10.load_data()

# Normalize inputs from 0-255 to 0.0-1.0
X_train = X_train.astype('float32') / 255.0
X_test = X_test.astype('float32') / 255.0

# One hot encode outputs
y_train = np_utils.to_categorical(y_train)
y_test = np_utils.to_categorical(y_test)
num_classes = y_test.shape[1]

# Create the 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)))
model.add(Conv2D(64, (3, 3), activation='relu', padding='same', kernel_constraint=maxnorm(3)))
model.add(Dropout(0.2))
model.add(Conv2D(64, (3, 3), activation='relu', padding='same', kernel_constraint=maxnorm(3)))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Conv2D(128, (3, 3), activation='relu', padding='same', kernel_constraint=maxnorm(3)))
model.add(Dropout(0.2))
model.add(Conv2D(128, (3, 3), activation='relu', padding='same', kernel_constraint=maxnorm(3)))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Flatten())
model.add(Dropout(0.2))
model.add(Dense(1024, 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'))

# Compile model
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())
```

```
# Fit the model
history = model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=epochs, batch_size=32)

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

```
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  (None, 16, 16, 32)       0
 2D)

 conv2d_4 (Conv2D)           (None, 16, 16, 64)        18496

 dropout_3 (Dropout)         (None, 16, 16, 64)        0

 conv2d_5 (Conv2D)           (None, 16, 16, 64)        36928

 max_pooling2d_2 (MaxPooling  (None, 8, 8, 64)         0
 2D)

 conv2d_6 (Conv2D)           (None, 8, 8, 128)         73856

 dropout_4 (Dropout)         (None, 8, 8, 128)         0

 conv2d_7 (Conv2D)           (None, 8, 8, 128)         147584

 max_pooling2d_3 (MaxPooling  (None, 4, 4, 128)        0
 2D)

 flatten_1 (Flatten)         (None, 2048)              0

 dropout_5 (Dropout)         (None, 2048)              0

 dense_2 (Dense)             (None, 1024)              2098176

 dropout_6 (Dropout)         (None, 1024)              0

 dense_3 (Dense)             (None, 512)               524800

 dropout_7 (Dropout)         (None, 512)               0

 dense_4 (Dense)             (None, 10)                5130

=================================================================
Total params: 2,915,114
Trainable params: 2,915,114
Non-trainable params: 0
_________________________________________________________________
/usr/local/lib/python3.10/dist-packages/keras/optimizers/legacy/gradient_descent.py:114: UserWarning: The `lr` argument is deprecated, use
  super().__init__(name, **kwargs)
None
Epoch 1/5
1563/1563 [==============================] - 342s 218ms/step - loss: 1.9313 - accuracy: 0.2843 - val_loss: 1.7142 - val_accuracy: 0.3886
Epoch 2/5
1563/1563 [==============================] - 345s 221ms/step - loss: 1.5361 - accuracy: 0.4434 - val_loss: 1.4258 - val_accuracy: 0.4928
Epoch 3/5
```

```
import numpy
# Predict the first 4 images of the test data
predictions = model.predict(X_test[:4])
# Convert the predictions to class labels
predicted_labels = numpy.argmax(predictions, axis=1)
# Convert the actual labels to class labels
actual_labels = numpy.argmax(y_test[:4], axis=1)

# Print the predicted and actual labels for the first 4 images
print("Predicted labels:", predicted_labels)
print("Actual labels:   ", actual_labels)
```

```
1/1 [==============================] - 0s 202ms/step
Predicted labels: [3 1 8 0]
Actual labels:    [3 8 8 0]
```

```
import matplotlib.pyplot as plt

# Plot the training and validation loss
```

```
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('Model Loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['train', 'val'], loc='upper right')
plt.show()

# Plot the training and validation accuracy
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('Model Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(['train', 'val'], loc='lower right')
plt.show()
```