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In [3]:
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import numpy as np
import os
import cv2
import matplotlib.pyplot as plt
from keras.models import Sequential
from keras.layers import Dense
from keras.layers import Dropout
from keras.layers import Flatten
from keras.layers.convolutional import Conv2D
from keras.layers.convolutional import MaxPooling2D
from keras.optimizers import Adam
from keras.utils.np_utils import to categorical
from sklearn.model_selection import train test split
from keras.preprocessing.image import ImageDataGenerator
myData = "C:/Users/ASUS/Downloads/Deep learning/Szakdolgozat program/myData"
batch size val=20
steps_per_epoch_val=2000
epochs_val=10
imageDimension = (32,32)
testRate = 0.2
validationRate = 0.2
count = 0
images = []
classNo = []
myList = os.listdir(myData)
print("Total Classes Detected:",len(myList))
noOfClasses=len(myList)
for x in range (0,len(myList)):
   myPicList = os.listdir(myData+"/"+str(count))
    for y in myPicList:
        curImg = cv2.imread(myData+"/"+str(count)+"/"+y)
       images.append(curImg)
       classNo.append(count)
   print(count, end =" ")
    count +=1
print(" ")
images = np.array(images)
classNo = np.array(classNo)
print(count)
X_train, X_test, y_train, y_test = train_test_split(images,
                                                     test size=testRate)
X train, X validation, y train, y validation = train test split(X train,
                                                                 v train,
                                                                 test size=validationRate)
def bgr2gray(img):
   img = cv2.cvtColor(img,cv2.COLOR_BGR2GRAY)
   return img
def equalizeHistogram(img):
   img =cv2.equalizeHist(img)
   return img
def preprocess(img):
   img = bgr2gray(img)
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img=equalizeHistogram(img)
    img = img/255
    return img
X train=np.array(list(map(preprocess, X train)))
X validation=np.array(list(map(preprocess, X validation)))
X_test=np.array(list(map(preprocess,X_test)))
X train=X train.reshape(X train.shape[0], X train.shape[1], X train.shape[2], 1)
X validation=X validation.reshape(X validation.shape[0], X validation.shape[1], X validation.shape[2
],1)
X test=X test.reshape(X test.shape[0], X test.shape[1], X test.shape[2], 1)
dataGen=ImageDataGenerator(width shift range=0.1,
                            height shift range=0.1,
                            zoom_range=0.2,
                            shear range=0.1,
                            rotation range=10)
dataGen.fit(X train)
y train = to categorical(y train,
                         noOfClasses)
y_validation = to_categorical(y_validation,
                             noOfClasses)
y test = to categorical(y test,
                       noOfClasses)
def myModel():
   model= Sequential()
   model.add((Conv2D(60,(3,3),input_shape=(imageDimension[0],
                                           imageDimension[1],
                                            1),
                                        activation='relu')))
   model.add((Conv2D(60, (3,3), activation='relu')))
    model.add(MaxPooling2D(pool size=(2,2)))
    model.add((Conv2D(30, (3,3), activation='relu')))
   model.add(MaxPooling2D(pool_size=(2,2)))
   model.add(Dropout(0.5))
   model.add(Flatten())
   model.add(Dense(500,activation='relu'))
   model.add(Dropout(0.5))
   model.add(Dense(noOfClasses,activation='softmax'))
   model.compile(Adam(lr=0.001),loss='categorical_crossentropy',metrics=['accuracy'])
    return model
model = myModel()
print(model.summary())
history=model.fit_generator(dataGen.flow(X_train,
             y train,
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batch size=batch size val),
                                                   steps_per_epoch=steps_per_epoch_val,
                                                   epochs=epochs val,
                                                   validation data=(X validation,
                                                                                y_validation))
plt.figure(1)
plt.plot(history.history['loss'])
plt.plot(history.history['val loss'])
plt.legend(['training','validation'])
plt.title('loss')
plt.xlabel('epoch')
plt.figure(2)
plt.plot(history.history['accuracy'])
plt.plot(history.history['val accuracy'])
plt.legend(['training','validation'])
plt.title('Acurracy')
plt.xlabel('epoch')
plt.show()
score =model.evaluate(X_test,y_test,verbose=0)
print('Test loss:',score[0])
print('Test Accuracy:',score[1])
Total Classes Detected: 43
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6 37 38 39 40 41 42
Model: "sequential_3"
Layer (type)
                                                   Output Shape
                                                                                                    Param #
______
conv2d 7 (Conv2D)
                                                    (None, 30, 30, 60)
                                                                                                  600
conv2d 8 (Conv2D)
                                                     (None, 28, 28, 60)
                                                                                                    32460
max pooling2d 5 (MaxPooling2 (None, 14, 14, 60)
conv2d 9 (Conv2D)
                                                     (None, 12, 12, 30)
                                                                                                    16230
max pooling2d 6 (MaxPooling2 (None, 6, 6, 30)
dropout 5 (Dropout)
                                                     (None, 6, 6, 30)
flatten 3 (Flatten)
                                                     (None, 1080)
dense 5 (Dense)
                                                     (None, 500)
                                                                                                    540500
                                                     (None, 500)
dropout 6 (Dropout)
                                                                                                    21543
dense_6 (Dense)
                                                     (None, 43)
Total params: 611,333
Trainable params: 611,333
Non-trainable params: 0
None
Epoch 1/10
al_loss: 0.5513 - val_accuracy: 0.8360
Epoch 2/10
2000/2000 [=========== ] - 209s 104ms/step - loss: 0.9397 - accuracy: 0.6980 - v
al loss: 0.2071 - val accuracy: 0.9585
Epoch 3/10
al loss: 0.1148 - val accuracy: 0.9738
Epoch 4/10
2000/2000 [============== ] - 213s 106ms/step - loss: 0.5655 - accuracy: 0.8173 - v
al loss: 0.1010 - val accuracy: 0.9810
Epoch 5/10
2000/2000 [============== ] - 215s 107ms/step - loss: 0.4915 - accuracy: 0.8442 - v
al loss: 0.0558 - val accuracy: 0.9881
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Epoch 6/10
2000/2000 [============== ] - 218s 109ms/step - loss: 0.4461 - accuracy: 0.8559 - v
al loss: 0.0546 - val accuracy: 0.9878
Epoch 7/10
al loss: 0.0491 - val accuracy: 0.9889
Epoch 8/10
2000/2000 [============= ] - 217s 108ms/step - loss: 0.3853 - accuracy: 0.8780 - v
al loss: 0.0602 - val accuracy: 0.9885
Epoch 9/10
2000/2000 [============== ] - 219s 109ms/step - loss: 0.3690 - accuracy: 0.8836 - v
al loss: 0.0377 - val accuracy: 0.9894
Epoch 10/10
2000/2000 [========== ] - 219s 109ms/step - loss: 0.3507 - accuracy: 0.8887 - v
al loss: 0.0345 - val accuracy: 0.9914
4
                    loss
2.00
                                 training
1.75
                                 validation
1.50
1.25
1.00
0.75
0.50
 0.25
 0.00
            'n
                                  8
                          6
                    epoch
                  Acurracy
1.0
0.9
0.8
 0.7
 0.6
 0.5
                                training
                                validation
           ż
                  4
    Ó
                   epoch
Test loss: 0.036753648272769036
Test Accuracy: 0.9916666746139526
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