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In [3]: import numpy as np
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout, BatchNormalization
from tensorflow.keras.optimizers import Adam
from sklearn.model_selection import train_test_split
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from skimage.transform import resize
from tensorflow.keras.regularizers import l2

def loadingData():
    data_full = np.load('data_train.npy')
    labels_all = np.load('labels_train.npy')

    #Reshape and Resize data
    reshaped_data = data.T.reshape((-1, 300, 300, 3))
    resizing_data = np.array([resize(img, (100, 100, 3), anti_aliasing=True) for img in reshaped_data])

    # Preprocessing the data
    normalizingData = resizing_data / 255.0

    return normalizingData, labels_all

def building_model():
    model = Sequential([
        Conv2D(64, (3, 3), activation='relu', input_shape=(100, 100, 3)),
        BatchNormalization(),
        MaxPooling2D((2, 2)),

        Conv2D(128, (3, 3), activation='relu'),
        BatchNormalization(),
        MaxPooling2D((2, 2)),

        Conv2D(256, (3, 3), activation='relu'),
        BatchNormalization(),
        MaxPooling2D((2, 2)),

        Conv2D(512, (3, 3), activation='relu'),
        BatchNormalization(),
        MaxPooling2D((2, 2)),

        #Increasing Complexity of the model and adjusting dropout rate with more Dense
        Flatten(),
        Dense(1024, activation='relu', kernel_regularizer=l2(0.001)),
        Dropout(0.4),
        Dense(512, activation='relu', kernel_regularizer=l2(0.001)),
        Dropout(0.3),
        Dense(9, activation='softmax')
    ])

    return model

def trainingSet():
    data_full, labels_all = loadingData()

    # Using Data augmentation generator and increasing it.
    datagen = ImageDataGenerator(
        rotation_range=20,
        width_shift_range=0.15,
        height_shift_range=0.15,
        zoom_range=0.2,
        horizontal_flip=True,
        fill_mode='nearest'
    )

    # Splitting the data into 80% of Training data and 20% Testing data.
    train_data, val_data, train_labels, val_labels = train_test_split(data_full, labels_all, test_size=0.2, random_state=42)

    model = building_model()
    model.compile(optimizer=Adam(learning_rate=0.0003), loss='sparse_categorical_crossentropy', metrics=['accuracy'])

    # Callbacks which keeps an eye on validation_loss:
    earlyStopping = tf.keras.callbacks.EarlyStopping(monitor='val_loss', patience=15)
    learningRate_scheduler = tf.keras.callbacks.ReduceLROnPlateau(monitor='val_loss', factor=0.1, patience=7, min_lr=0.00001)
    modellingCheckpoint = tf.keras.callbacks.ModelCheckpoint('my_best_model.h5', save_best_only=True)

    # Training the model with the parameters given in the callbacks functions.
    trainingGenerator = datagen.flow(train_data, train_labels, batch_size=32)
    model.fit(trainingGenerator, epochs=60, validation_data=(val_data, val_labels),
            callbacks=[earlyStopping, learningRate_scheduler, modellingCheckpoint])

if __name__ == "__main__":
    trainingSet()

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Epoch 1/60  
212/212 [=====] - 12s 52ms/step - loss: 5.2616 - accuracy: 0.1220 - val\_loss: 6.0595 - val\_accuracy: 0.1113 - lr: 3.0000e-04

Epoch 2/60  
212/212 [=====] - 11s 50ms/step - loss: 4.6421 - accuracy: 0.1334 - val\_loss: 4.8237 - val\_accuracy: 0.1113 - lr: 3.0000e-04

Epoch 3/60  
212/212 [=====] - 11s 50ms/step - loss: 4.3487 - accuracy: 0.1516 - val\_loss: 4.1800 - val\_accuracy: 0.1551 - lr: 3.0000e-04

Epoch 4/60  
212/212 [=====] - 11s 50ms/step - loss: 4.0602 - accuracy: 0.1697 - val\_loss: 3.8687 - val\_accuracy: 0.2362 - lr: 3.0000e-04

Epoch 5/60  
212/212 [=====] - 11s 50ms/step - loss: 3.7566 - accuracy: 0.2039 - val\_loss: 3.7514 - val\_accuracy: 0.1652 - lr: 3.0000e-04

Epoch 6/60  
212/212 [=====] - 11s 50ms/step - loss: 3.4658 - accuracy: 0.2431 - val\_loss: 3.3578 - val\_accuracy: 0.2410 - lr: 3.0000e-04

Epoch 7/60  
212/212 [=====] - 11s 50ms/step - loss: 3.1677 - accuracy: 0.2936 - val\_loss: 3.0798 - val\_accuracy: 0.3102 - lr: 3.0000e-04

Epoch 8/60  
212/212 [=====] - 11s 50ms/step - loss: 2.8778 - accuracy: 0.3413 - val\_loss: 2.9942 - val\_accuracy: 0.2747 - lr: 3.0000e-04

Epoch 9/60  
212/212 [=====] - 11s 51ms/step - loss: 2.6133 - accuracy: 0.4058 - val\_loss: 2.3408 - val\_accuracy: 0.5145 - lr: 3.0000e-04

Epoch 10/60  
212/212 [=====] - 11s 50ms/step - loss: 2.3825 - accuracy: 0.4563 - val\_loss: 2.2402 - val\_accuracy: 0.5631 - lr: 3.0000e-04

Epoch 11/60  
212/212 [=====] - 11s 50ms/step - loss: 2.1965 - accuracy: 0.5102 - val\_loss: 2.0304 - val\_accuracy: 0.5583 - lr: 3.0000e-04

Epoch 12/60  
212/212 [=====] - 11s 51ms/step - loss: 2.0270 - accuracy: 0.5548 - val\_loss: 1.8586 - val\_accuracy: 0.6264 - lr: 3.0000e-04

Epoch 13/60  
212/212 [=====] - 11s 50ms/step - loss: 1.9823 - accuracy: 0.5791 - val\_loss: 1.8519 - val\_accuracy: 0.6128 - lr: 3.0000e-04

Epoch 14/60  
212/212 [=====] - 11s 50ms/step - loss: 1.8481 - accuracy: 0.6183 - val\_loss: 1.8244 - val\_accuracy: 0.6507 - lr: 3.0000e-04

Epoch 15/60  
212/212 [=====] - 11s 50ms/step - loss: 1.7489 - accuracy: 0.6515 - val\_loss: 1.7540 - val\_accuracy: 0.6696 - lr: 3.0000e-04

Epoch 16/60  
212/212 [=====] - 10s 49ms/step - loss: 1.6674 - accuracy: 0.6846 - val\_loss: 1.7666 - val\_accuracy: 0.6459 - lr: 3.0000e-04

Epoch 17/60  
212/212 [=====] - 11s 50ms/step - loss: 1.6143 - accuracy: 0.6980 - val\_loss: 1.5134 - val\_accuracy: 0.7217 - lr: 3.0000e-04

Epoch 18/60  
212/212 [=====] - 10s 49ms/step - loss: 1.5818 - accuracy: 0.7154 - val\_loss: 1.5670 - val\_accuracy: 0.7294 - lr: 3.0000e-04

Epoch 19/60  
212/212 [=====] - 11s 50ms/step - loss: 1.5035 - accuracy: 0.7384 - val\_loss: 1.3460 - val\_accuracy: 0.8046 - lr: 3.0000e-04

Epoch 20/60  
212/212 [=====] - 11s 51ms/step - loss: 1.4508 - accuracy: 0.7572 - val\_loss: 1.3157 - val\_accuracy: 0.8022 - lr: 3.0000e-04

Epoch 21/60  
212/212 [=====] - 10s 49ms/step - loss: 1.4308 - accuracy: 0.7640 - val\_loss: 1.6047 - val\_accuracy: 0.7235 - lr: 3.0000e-04

Epoch 22/60  
212/212 [=====] - 10s 49ms/step - loss: 1.3760 - accuracy: 0.7760 - val\_loss: 1.3597 - val\_accuracy: 0.7993 - lr: 3.0000e-04

Epoch 23/60  
212/212 [=====] - 11s 50ms/step - loss: 1.3248 - accuracy: 0.7936 - val\_loss: 1.2839 - val\_accuracy: 0.8058 - lr: 3.0000e-04

Epoch 24/60  
212/212 [=====] - 10s 49ms/step - loss: 1.3269 - accuracy: 0.7948 - val\_loss: 1.8637 - val\_accuracy: 0.5986 - lr: 3.0000e-04

Epoch 25/60  
212/212 [=====] - 10s 49ms/step - loss: 1.3179 - accuracy: 0.8016 - val\_loss: 1.4266 - val\_accuracy: 0.7880 - lr: 3.0000e-04

Epoch 26/60  
212/212 [=====] - 10s 49ms/step - loss: 1.2873 - accuracy: 0.8035 - val\_loss: 1.6391 - val\_accuracy: 0.6951 - lr: 3.0000e-04

Epoch 27/60  
212/212 [=====] - 11s 50ms/step - loss: 1.2651 - accuracy: 0.8117 - val\_loss: 1.1491 - val\_accuracy: 0.8632 - lr: 3.0000e-04

Epoch 28/60  
212/212 [=====] - 10s 49ms/step - loss: 1.2296 - accuracy: 0.8244 - val\_loss: 1.1802 - val\_accuracy: 0.8289 - lr: 3.0000e-04

Epoch 29/60  
212/212 [=====] - 11s 50ms/step - loss: 1.2015 - accuracy: 0.8333 - val\_loss: 1.0767 - val\_accuracy: 0.8727 - lr: 3.0000e-04

Epoch 30/60  
212/212 [=====] - 10s 49ms/step - loss: 1.1904 - accuracy: 0.8297 - val\_loss: 1.1349 - val\_accuracy: 0.8514 - lr: 3.0000e-04

Epoch 31/60  
212/212 [=====] - 10s 49ms/step - loss: 1.1263 - accuracy: 0.8519 - val\_loss: 1.3298 - val\_accuracy: 0.

7809 - lr: 3.0000e-04  
Epoch 32/60  
212/212 [=====] - 11s 50ms/step - loss: 1.1517 - accuracy: 0.8342 - val\_loss: 1.1861 - val\_accuracy: 0.  
8336 - lr: 3.0000e-04  
Epoch 33/60  
212/212 [=====] - 10s 49ms/step - loss: 1.1115 - accuracy: 0.8496 - val\_loss: 1.1264 - val\_accuracy: 0.  
8425 - lr: 3.0000e-04  
Epoch 34/60  
212/212 [=====] - 10s 49ms/step - loss: 1.0904 - accuracy: 0.8580 - val\_loss: 1.1147 - val\_accuracy: 0.  
8455 - lr: 3.0000e-04  
Epoch 35/60  
212/212 [=====] - 11s 50ms/step - loss: 1.0683 - accuracy: 0.8592 - val\_loss: 1.0236 - val\_accuracy: 0.  
8769 - lr: 3.0000e-04  
Epoch 36/60  
212/212 [=====] - 10s 49ms/step - loss: 1.0630 - accuracy: 0.8579 - val\_loss: 1.0774 - val\_accuracy: 0.  
8555 - lr: 3.0000e-04  
Epoch 37/60  
212/212 [=====] - 11s 50ms/step - loss: 1.0485 - accuracy: 0.8617 - val\_loss: 1.0362 - val\_accuracy: 0.  
8733 - lr: 3.0000e-04  
Epoch 38/60  
212/212 [=====] - 10s 49ms/step - loss: 1.0220 - accuracy: 0.8707 - val\_loss: 4.6540 - val\_accuracy: 0.  
3102 - lr: 3.0000e-04  
Epoch 39/60  
212/212 [=====] - 10s 49ms/step - loss: 1.0268 - accuracy: 0.8715 - val\_loss: 1.1129 - val\_accuracy: 0.  
8591 - lr: 3.0000e-04  
Epoch 40/60  
212/212 [=====] - 11s 50ms/step - loss: 1.0137 - accuracy: 0.8724 - val\_loss: 1.0895 - val\_accuracy: 0.  
8496 - lr: 3.0000e-04  
Epoch 41/60  
212/212 [=====] - 11s 50ms/step - loss: 0.9910 - accuracy: 0.8798 - val\_loss: 0.9729 - val\_accuracy: 0.  
8893 - lr: 3.0000e-04  
Epoch 42/60  
212/212 [=====] - 10s 49ms/step - loss: 1.0171 - accuracy: 0.8736 - val\_loss: 6.2417 - val\_accuracy: 0.  
1208 - lr: 3.0000e-04  
Epoch 43/60  
212/212 [=====] - 10s 49ms/step - loss: 0.9588 - accuracy: 0.8867 - val\_loss: 2.0139 - val\_accuracy: 0.  
5204 - lr: 3.0000e-04  
Epoch 44/60  
212/212 [=====] - 10s 49ms/step - loss: 1.0101 - accuracy: 0.8724 - val\_loss: 1.0549 - val\_accuracy: 0.  
8680 - lr: 3.0000e-04  
Epoch 45/60  
212/212 [=====] - 11s 50ms/step - loss: 0.9778 - accuracy: 0.8746 - val\_loss: 5.4293 - val\_accuracy: 0.  
1658 - lr: 3.0000e-04  
Epoch 46/60  
212/212 [=====] - 10s 49ms/step - loss: 0.9600 - accuracy: 0.8793 - val\_loss: 1.0097 - val\_accuracy: 0.  
8739 - lr: 3.0000e-04  
Epoch 47/60  
212/212 [=====] - 10s 49ms/step - loss: 0.9565 - accuracy: 0.8838 - val\_loss: 4.6187 - val\_accuracy: 0.  
1308 - lr: 3.0000e-04  
Epoch 48/60  
212/212 [=====] - 10s 49ms/step - loss: 0.9330 - accuracy: 0.8857 - val\_loss: 1.2865 - val\_accuracy: 0.  
7762 - lr: 3.0000e-04  
Epoch 49/60  
212/212 [=====] - 11s 50ms/step - loss: 0.8760 - accuracy: 0.9045 - val\_loss: 0.8497 - val\_accuracy: 0.  
9207 - lr: 3.0000e-05  
Epoch 50/60  
212/212 [=====] - 11s 50ms/step - loss: 0.8383 - accuracy: 0.9152 - val\_loss: 0.8094 - val\_accuracy: 0.  
9325 - lr: 3.0000e-05  
Epoch 51/60  
212/212 [=====] - 11s 50ms/step - loss: 0.7968 - accuracy: 0.9218 - val\_loss: 0.7859 - val\_accuracy: 0.  
9307 - lr: 3.0000e-05  
Epoch 52/60  
212/212 [=====] - 11s 50ms/step - loss: 0.7874 - accuracy: 0.9227 - val\_loss: 0.7644 - val\_accuracy: 0.  
9361 - lr: 3.0000e-05  
Epoch 53/60  
212/212 [=====] - 10s 49ms/step - loss: 0.7541 - accuracy: 0.9286 - val\_loss: 0.7745 - val\_accuracy: 0.  
9278 - lr: 3.0000e-05  
Epoch 54/60  
212/212 [=====] - 11s 51ms/step - loss: 0.7408 - accuracy: 0.9329 - val\_loss: 0.7408 - val\_accuracy: 0.  
9355 - lr: 3.0000e-05  
Epoch 55/60  
212/212 [=====] - 11s 50ms/step - loss: 0.7208 - accuracy: 0.9335 - val\_loss: 0.7263 - val\_accuracy: 0.  
9384 - lr: 3.0000e-05  
Epoch 56/60  
212/212 [=====] - 11s 50ms/step - loss: 0.6948 - accuracy: 0.9350 - val\_loss: 0.7205 - val\_accuracy: 0.  
9366 - lr: 3.0000e-05  
Epoch 57/60  
212/212 [=====] - 11s 50ms/step - loss: 0.6939 - accuracy: 0.9357 - val\_loss: 0.7146 - val\_accuracy: 0.  
9372 - lr: 3.0000e-05  
Epoch 58/60  
212/212 [=====] - 11s 50ms/step - loss: 0.6878 - accuracy: 0.9372 - val\_loss: 0.6895 - val\_accuracy: 0.  
9384 - lr: 3.0000e-05  
Epoch 59/60  
212/212 [=====] - 11s 50ms/step - loss: 0.6715 - accuracy: 0.9346 - val\_loss: 0.6892 - val\_accuracy: 0.  
9390 - lr: 3.0000e-05  
Epoch 60/60  
212/212 [=====] - 11s 50ms/step - loss: 0.6571 - accuracy: 0.9350 - val\_loss: 0.6829 - val\_accuracy: 0.  
9337 - lr: 3.0000e-05