Pet Classification Model using CNN

DESCRIPTION

Project Objective: Build a CNN model that classifies the given pet images correctly into dog and cat images. The project scope document specifies the requirements for the project "Pet Classification Model Using CNN." Apart from specifying the functional and non-functional requirements for the project, it also serves as an input for project scoping.

Project Description and Scope: You are provided with a collection of images of pets, that is, cats and dogs. These images are of different sizes with varied lighting conditions and they should be used as inputs for your model.

You are expected to write the code for CNN image classification model using TensorFlow that trains on the data and calculates the accuracy score on the test data.

```
# Importing required libraries
In [1]:
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         from keras.preprocessing.image import load_img, ImageDataGenerator
         from tensorflow.keras.models import Sequential
         from tensorflow.keras.layers import InputLayer, Conv2D, MaxPooling2D, Dense, Droport
In [2]:
        import os
         os.getcwd()
Out[2]: '/content'
        # Loading test and train images
         data_gen = ImageDataGenerator()
         train images = data gen.flow from directory('/content/drive/MyDrive/datasets/data/
         test_images = data_gen.flow_from_directory('/content/drive/MyDrive/datasets/data/te
         Found 40 images belonging to 2 classes.
         Found 20 images belonging to 2 classes.
         image data = pd.DataFrame()
In [4]:
         image_data['train_images'] = train_images.filenames
         image_data['train_labels'] = train_images.classes
         image_data.head()
Out[4]:
           train_images train_labels
                                0
              dogs/1.jpg
         1
            dogs/10.jpg
                                0
         2
             dogs/11.jpg
                                0
            dogs/12.jpg
                                0
            dogs/13.jpg
                                0
```

In [5]: image_data.train_labels.value_counts()

Out[5]: 0 20 1 20

Name: train_labels, dtype: int64

In [6]: # View a sample of the data
import random
cat_sample_image = image_data[image_data.train_labels==1].sample(1).values[0][0]
load_img("/content/drive/MyDrive/datasets/data/train/" + cat_sample_image)



Out[7]:



```
# Building the CNN model
In [8]:
        model = Sequential([
                             InputLayer(input_shape=(256,256,3)),
                             Conv2D(32, (5,5), activation='relu'),
                             BatchNormalization(),
                            MaxPooling2D(pool_size=(2,2), strides=2),
                            Conv2D(64, (5,5), activation='relu'),
                             BatchNormalization(),
                            MaxPooling2D(pool_size=(2,2), strides=2),
                             Flatten(),
                            Dense(32, activation='relu'),
                            Dropout(0.4),
                             BatchNormalization(),
                            Dense(2, activation='softmax')
        ])
        model.compile(loss='categorical_crossentropy', optimizer='rmsprop', metrics=['accur
        model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 252, 252, 32)	2432
<pre>batch_normalization (BatchN ormalization)</pre>	(None, 252, 252, 32)	128
<pre>max_pooling2d (MaxPooling2D)</pre>	(None, 126, 126, 32)	0
conv2d_1 (Conv2D)	(None, 122, 122, 64)	51264
<pre>batch_normalization_1 (Batc hNormalization)</pre>	(None, 122, 122, 64)	256
<pre>max_pooling2d_1 (MaxPooling 2D)</pre>	(None, 61, 61, 64)	0
flatten (Flatten)	(None, 238144)	0
dense (Dense)	(None, 32)	7620640
dropout (Dropout)	(None, 32)	0
<pre>batch_normalization_2 (Batc hNormalization)</pre>	(None, 32)	128
dense_1 (Dense)	(None, 2)	66

Total params: 7,674,914 Trainable params: 7,674,658 Non-trainable params: 256

In [9]: history = model.fit(train_images, epochs=200, batch_size=10, validation_data=test_:

```
Epoch 1/200
0 - val loss: 1471.1606 - val accuracy: 0.5000
Epoch 2/200
7750 - val_loss: 36.1675 - val_accuracy: 0.4000
Epoch 3/200
2/2 [============= ] - 1s 756ms/step - loss: 0.3812 - accuracy: 0.
8000 - val_loss: 21.9132 - val_accuracy: 0.3000
Epoch 4/200
2/2 [================ ] - 1s 280ms/step - loss: 0.2405 - accuracy: 0.
9250 - val_loss: 16.5973 - val_accuracy: 0.3500
Epoch 5/200
2/2 [============== ] - 1s 266ms/step - loss: 0.2332 - accuracy: 0.
9000 - val_loss: 22.7031 - val_accuracy: 0.3500
Epoch 6/200
2/2 [==============] - 1s 406ms/step - loss: 0.2313 - accuracy: 0.
9000 - val_loss: 26.6634 - val_accuracy: 0.4000
Epoch 7/200
2/2 [================ ] - 1s 265ms/step - loss: 0.1200 - accuracy: 0.
9750 - val_loss: 15.8650 - val_accuracy: 0.4000
Epoch 8/200
2/2 [================ ] - 1s 785ms/step - loss: 0.0858 - accuracy: 1.
0000 - val_loss: 11.7850 - val_accuracy: 0.4000
Epoch 9/200
2/2 [=============== ] - 1s 387ms/step - loss: 0.0931 - accuracy: 1.
0000 - val_loss: 12.9374 - val_accuracy: 0.4000
Epoch 10/200
2/2 [=============== ] - 1s 843ms/step - loss: 0.1046 - accuracy: 0.
9750 - val_loss: 9.0130 - val_accuracy: 0.4500
Epoch 11/200
2/2 [============= ] - 1s 352ms/step - loss: 0.1170 - accuracy: 0.
9750 - val_loss: 11.4859 - val_accuracy: 0.4000
Epoch 12/200
2/2 [==============] - 0s 342ms/step - loss: 0.0662 - accuracy: 1.
0000 - val_loss: 10.3335 - val_accuracy: 0.4000
Epoch 13/200
2/2 [=============== ] - 0s 361ms/step - loss: 0.0648 - accuracy: 1.
0000 - val loss: 7.5905 - val accuracy: 0.4500
Epoch 14/200
2/2 [=============== ] - 0s 348ms/step - loss: 0.0701 - accuracy: 0.
9750 - val_loss: 8.1267 - val_accuracy: 0.4000
Epoch 15/200
2/2 [================ ] - 0s 182ms/step - loss: 0.0934 - accuracy: 0.
9750 - val_loss: 8.5931 - val_accuracy: 0.4500
Epoch 16/200
0000 - val loss: 7.1657 - val accuracy: 0.4500
Epoch 17/200
0000 - val_loss: 4.6559 - val_accuracy: 0.3500
Epoch 18/200
2/2 [================ ] - 0s 364ms/step - loss: 0.0747 - accuracy: 1.
0000 - val_loss: 4.4573 - val_accuracy: 0.3500
Epoch 19/200
0000 - val loss: 6.9000 - val accuracy: 0.4500
Epoch 20/200
2/2 [============= ] - 0s 171ms/step - loss: 0.0616 - accuracy: 1.
0000 - val loss: 1.9438 - val accuracy: 0.3500
Epoch 21/200
2/2 [=============] - 0s 164ms/step - loss: 0.0533 - accuracy: 1.
0000 - val_loss: 7.5754 - val_accuracy: 0.4500
Epoch 22/200
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2/2 [============= ] - 0s 362ms/step - loss: 0.0541 - accuracy: 1.
0000 - val_loss: 4.1339 - val_accuracy: 0.5000
Epoch 23/200
9750 - val loss: 4.8918 - val accuracy: 0.5000
Epoch 24/200
2/2 [==============] - 0s 373ms/step - loss: 0.0356 - accuracy: 1.
0000 - val_loss: 4.3631 - val_accuracy: 0.5000
Epoch 25/200
2/2 [=============== ] - 0s 169ms/step - loss: 0.0644 - accuracy: 0.
9750 - val_loss: 3.6943 - val_accuracy: 0.4000
Epoch 26/200
2/2 [============ ] - 0s 172ms/step - loss: 0.0388 - accuracy: 1.
0000 - val loss: 3.3079 - val accuracy: 0.4500
Epoch 27/200
2/2 [==============] - 0s 360ms/step - loss: 0.0534 - accuracy: 1.
0000 - val_loss: 2.5599 - val_accuracy: 0.5000
Epoch 28/200
2/2 [================ ] - 0s 165ms/step - loss: 0.0396 - accuracy: 1.
0000 - val_loss: 2.7288 - val_accuracy: 0.5000
Epoch 29/200
2/2 [============= ] - 0s 360ms/step - loss: 0.1080 - accuracy: 0.
9750 - val_loss: 2.3251 - val_accuracy: 0.4500
Epoch 30/200
2/2 [==============] - 0s 165ms/step - loss: 0.0426 - accuracy: 1.
0000 - val loss: 2.5709 - val accuracy: 0.4000
Epoch 31/200
2/2 [================ ] - 0s 349ms/step - loss: 0.0379 - accuracy: 1.
0000 - val_loss: 2.1288 - val_accuracy: 0.4000
Epoch 32/200
2/2 [============ ] - 0s 341ms/step - loss: 0.0577 - accuracy: 1.
0000 - val_loss: 2.5683 - val_accuracy: 0.5000
Epoch 33/200
2/2 [==============] - 0s 167ms/step - loss: 0.0384 - accuracy: 1.
0000 - val_loss: 2.2442 - val_accuracy: 0.5500
Epoch 34/200
2/2 [================ ] - 0s 362ms/step - loss: 0.0508 - accuracy: 1.
0000 - val_loss: 1.8701 - val_accuracy: 0.5500
Epoch 35/200
2/2 [============ ] - 0s 348ms/step - loss: 0.0360 - accuracy: 1.
0000 - val_loss: 1.8815 - val_accuracy: 0.5500
Epoch 36/200
2/2 [============= ] - 0s 169ms/step - loss: 0.0454 - accuracy: 0.
9750 - val_loss: 2.0963 - val_accuracy: 0.4500
Epoch 37/200
2/2 [================= ] - 0s 165ms/step - loss: 0.0355 - accuracy: 1.
0000 - val_loss: 2.2055 - val_accuracy: 0.5000
Epoch 38/200
2/2 [============= ] - 0s 356ms/step - loss: 0.0369 - accuracy: 1.
0000 - val loss: 1.6505 - val accuracy: 0.5000
Epoch 39/200
2/2 [================ ] - 0s 343ms/step - loss: 0.0374 - accuracy: 1.
0000 - val_loss: 1.9041 - val_accuracy: 0.5500
Epoch 40/200
2/2 [============] - 0s 173ms/step - loss: 0.0576 - accuracy: 0.
9750 - val_loss: 5.0113 - val_accuracy: 0.4500
Epoch 41/200
2/2 [============== ] - 0s 165ms/step - loss: 0.0745 - accuracy: 0.
9750 - val loss: 5.2989 - val accuracy: 0.6000
2/2 [================ ] - 0s 354ms/step - loss: 0.0661 - accuracy: 0.
9750 - val_loss: 4.4069 - val_accuracy: 0.4500
Epoch 43/200
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9750 - val_loss: 4.2989 - val_accuracy: 0.4000
Epoch 44/200
2/2 [============ ] - 0s 358ms/step - loss: 0.0567 - accuracy: 1.
0000 - val_loss: 4.6008 - val_accuracy: 0.3000
Epoch 45/200
2/2 [============= ] - 0s 201ms/step - loss: 0.0480 - accuracy: 1.
0000 - val_loss: 3.5907 - val_accuracy: 0.3500
Epoch 46/200
2/2 [==============] - 0s 172ms/step - loss: 0.0352 - accuracy: 1.
0000 - val_loss: 3.2444 - val_accuracy: 0.3500
Epoch 47/200
2/2 [==============] - 0s 167ms/step - loss: 0.0192 - accuracy: 1.
0000 - val loss: 3.9659 - val accuracy: 0.3500
Epoch 48/200
2/2 [=============== ] - 0s 170ms/step - loss: 0.0242 - accuracy: 1.
0000 - val_loss: 3.2751 - val_accuracy: 0.4500
Epoch 49/200
2/2 [=============== ] - 0s 163ms/step - loss: 0.0414 - accuracy: 1.
0000 - val_loss: 2.1342 - val_accuracy: 0.5000
Epoch 50/200
2/2 [=============== ] - 0s 168ms/step - loss: 0.0565 - accuracy: 0.
9750 - val_loss: 5.6486 - val_accuracy: 0.3500
Epoch 51/200
2/2 [=============== ] - 0s 167ms/step - loss: 0.0402 - accuracy: 1.
0000 - val_loss: 4.9844 - val_accuracy: 0.4000
Epoch 52/200
2/2 [==============] - 0s 358ms/step - loss: 0.0508 - accuracy: 1.
0000 - val_loss: 4.0055 - val_accuracy: 0.4000
Epoch 53/200
0000 - val loss: 4.1390 - val accuracy: 0.4000
Epoch 54/200
2/2 [=============== ] - 0s 342ms/step - loss: 0.0354 - accuracy: 1.
0000 - val_loss: 4.1091 - val_accuracy: 0.5000
Epoch 55/200
2/2 [================ ] - 0s 351ms/step - loss: 0.0338 - accuracy: 1.
0000 - val_loss: 3.4925 - val_accuracy: 0.5000
Epoch 56/200
2/2 [============= ] - 0s 345ms/step - loss: 0.0397 - accuracy: 1.
0000 - val_loss: 3.1321 - val_accuracy: 0.5000
Epoch 57/200
2/2 [============= ] - 0s 178ms/step - loss: 0.0403 - accuracy: 1.
0000 - val_loss: 2.3907 - val_accuracy: 0.4500
Epoch 58/200
9750 - val_loss: 2.4176 - val_accuracy: 0.5000
Epoch 59/200
2/2 [============= ] - 0s 164ms/step - loss: 0.0426 - accuracy: 1.
0000 - val loss: 2.1207 - val accuracy: 0.5000
Epoch 60/200
2/2 [==============] - 0s 350ms/step - loss: 0.0242 - accuracy: 1.
0000 - val_loss: 2.1822 - val_accuracy: 0.4500
Epoch 61/200
2/2 [=================== ] - 0s 183ms/step - loss: 0.0160 - accuracy: 1.
0000 - val_loss: 2.3000 - val_accuracy: 0.4500
Epoch 62/200
2/2 [============== ] - 0s 360ms/step - loss: 0.0447 - accuracy: 0.
9750 - val loss: 3.2858 - val accuracy: 0.4500
Epoch 63/200
0000 - val loss: 2.7726 - val accuracy: 0.4000
Epoch 64/200
2/2 [=================== ] - 0s 352ms/step - loss: 0.0304 - accuracy: 1.
0000 - val_loss: 2.4709 - val_accuracy: 0.4000
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Epoch 65/200
2/2 [============= ] - 0s 164ms/step - loss: 0.0140 - accuracy: 1.
0000 - val loss: 2.1004 - val accuracy: 0.3000
Epoch 66/200
2/2 [==============] - 0s 166ms/step - loss: 0.0183 - accuracy: 1.
0000 - val_loss: 2.0089 - val_accuracy: 0.4000
Epoch 67/200
0000 - val_loss: 2.2072 - val_accuracy: 0.4000
Epoch 68/200
2/2 [================ ] - 0s 354ms/step - loss: 0.0115 - accuracy: 1.
0000 - val_loss: 2.4151 - val_accuracy: 0.3500
Epoch 69/200
2/2 [============ ] - 0s 178ms/step - loss: 0.0136 - accuracy: 1.
0000 - val_loss: 2.2151 - val_accuracy: 0.4500
Epoch 70/200
2/2 [==============] - 0s 172ms/step - loss: 0.0249 - accuracy: 1.
0000 - val_loss: 1.8448 - val_accuracy: 0.4500
Epoch 71/200
2/2 [================ ] - 0s 175ms/step - loss: 0.0189 - accuracy: 1.
0000 - val_loss: 1.8355 - val_accuracy: 0.5500
Epoch 72/200
2/2 [=============== ] - 0s 346ms/step - loss: 0.0247 - accuracy: 1.
0000 - val_loss: 2.0115 - val_accuracy: 0.4000
Epoch 73/200
2/2 [=============== ] - 0s 349ms/step - loss: 0.0391 - accuracy: 1.
0000 - val_loss: 1.5961 - val_accuracy: 0.4000
Epoch 74/200
2/2 [=============== ] - 0s 371ms/step - loss: 0.1109 - accuracy: 0.
9500 - val_loss: 3.0774 - val_accuracy: 0.4000
Epoch 75/200
2/2 [============ ] - 0s 351ms/step - loss: 0.0145 - accuracy: 1.
0000 - val_loss: 2.3169 - val_accuracy: 0.4500
Epoch 76/200
2/2 [==============] - 0s 358ms/step - loss: 0.0234 - accuracy: 1.
0000 - val_loss: 1.9271 - val_accuracy: 0.3500
Epoch 77/200
2/2 [=============== ] - 0s 378ms/step - loss: 0.0159 - accuracy: 1.
0000 - val_loss: 2.1458 - val_accuracy: 0.4500
Epoch 78/200
2/2 [=============== ] - 0s 358ms/step - loss: 0.0106 - accuracy: 1.
0000 - val_loss: 2.0432 - val_accuracy: 0.3500
Epoch 79/200
2/2 [================ ] - 0s 356ms/step - loss: 0.0259 - accuracy: 1.
0000 - val_loss: 1.9098 - val_accuracy: 0.5500
Epoch 80/200
0000 - val loss: 1.8839 - val accuracy: 0.5000
Epoch 81/200
0000 - val_loss: 2.4845 - val_accuracy: 0.3500
Epoch 82/200
2/2 [================ ] - 0s 364ms/step - loss: 0.0383 - accuracy: 1.
0000 - val_loss: 1.7034 - val_accuracy: 0.4000
Epoch 83/200
0000 - val loss: 1.7547 - val accuracy: 0.4500
Epoch 84/200
2/2 [============= ] - 0s 175ms/step - loss: 0.0132 - accuracy: 1.
0000 - val loss: 1.6087 - val accuracy: 0.4500
Epoch 85/200
2/2 [================= ] - 0s 177ms/step - loss: 0.0144 - accuracy: 1.
0000 - val_loss: 1.6425 - val_accuracy: 0.5000
Epoch 86/200
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2/2 [============= ] - 0s 354ms/step - loss: 0.0317 - accuracy: 1.
0000 - val loss: 1.4399 - val accuracy: 0.5000
Epoch 87/200
0000 - val loss: 1.5919 - val accuracy: 0.5500
Epoch 88/200
2/2 [==============] - 0s 169ms/step - loss: 0.0119 - accuracy: 1.
0000 - val_loss: 1.4299 - val_accuracy: 0.4500
Epoch 89/200
2/2 [=============== ] - 0s 168ms/step - loss: 0.0138 - accuracy: 1.
0000 - val_loss: 1.4517 - val_accuracy: 0.5000
Epoch 90/200
2/2 [============ ] - 0s 173ms/step - loss: 0.0100 - accuracy: 1.
0000 - val loss: 1.5922 - val accuracy: 0.5000
Epoch 91/200
2/2 [==============] - 0s 166ms/step - loss: 0.0173 - accuracy: 1.
0000 - val_loss: 1.3988 - val_accuracy: 0.4500
Epoch 92/200
2/2 [================ ] - 0s 172ms/step - loss: 0.0176 - accuracy: 1.
0000 - val_loss: 2.3366 - val_accuracy: 0.5500
Epoch 93/200
2/2 [============ ] - 0s 381ms/step - loss: 0.0226 - accuracy: 1.
0000 - val_loss: 1.6127 - val_accuracy: 0.6000
Epoch 94/200
2/2 [==============] - 0s 183ms/step - loss: 0.0188 - accuracy: 1.
0000 - val loss: 2.4432 - val accuracy: 0.5000
Epoch 95/200
2/2 [================ ] - 0s 366ms/step - loss: 0.0082 - accuracy: 1.
0000 - val_loss: 1.2472 - val_accuracy: 0.4500
Epoch 96/200
0000 - val_loss: 1.6641 - val_accuracy: 0.5000
Epoch 97/200
2/2 [=============== ] - 0s 356ms/step - loss: 0.0258 - accuracy: 1.
0000 - val_loss: 1.8794 - val_accuracy: 0.4500
Epoch 98/200
2/2 [================ ] - 0s 350ms/step - loss: 0.0103 - accuracy: 1.
0000 - val_loss: 1.6185 - val_accuracy: 0.5000
Epoch 99/200
2/2 [==============] - 0s 169ms/step - loss: 0.0074 - accuracy: 1.
0000 - val_loss: 1.6900 - val_accuracy: 0.5000
Epoch 100/200
2/2 [================ ] - 0s 167ms/step - loss: 0.0071 - accuracy: 1.
0000 - val_loss: 1.4471 - val_accuracy: 0.4500
Epoch 101/200
2/2 [================ ] - 0s 379ms/step - loss: 0.0117 - accuracy: 1.
0000 - val_loss: 1.3779 - val_accuracy: 0.4500
Epoch 102/200
2/2 [============= ] - 0s 170ms/step - loss: 0.0086 - accuracy: 1.
0000 - val loss: 2.5250 - val accuracy: 0.4500
Epoch 103/200
2/2 [================ ] - 0s 361ms/step - loss: 0.0117 - accuracy: 1.
0000 - val_loss: 1.6564 - val_accuracy: 0.4500
Epoch 104/200
2/2 [============] - 0s 360ms/step - loss: 0.0323 - accuracy: 1.
0000 - val loss: 1.9594 - val accuracy: 0.4500
Epoch 105/200
2/2 [============= ] - 0s 349ms/step - loss: 0.0267 - accuracy: 1.
0000 - val_loss: 1.2531 - val_accuracy: 0.5000
Epoch 106/200
2/2 [============= ] - 0s 166ms/step - loss: 0.0408 - accuracy: 1.
0000 - val_loss: 1.0532 - val_accuracy: 0.5000
Epoch 107/200
2/2 [==============] - 0s 168ms/step - loss: 0.0126 - accuracy: 1.
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0000 - val_loss: 1.0576 - val_accuracy: 0.5000
Epoch 108/200
2/2 [============= ] - 0s 174ms/step - loss: 0.0535 - accuracy: 0.
9750 - val_loss: 2.0351 - val_accuracy: 0.3500
Epoch 109/200
2/2 [============ ] - 0s 359ms/step - loss: 0.0090 - accuracy: 1.
0000 - val_loss: 1.7834 - val_accuracy: 0.3500
Epoch 110/200
2/2 [==============] - 0s 356ms/step - loss: 0.0132 - accuracy: 1.
0000 - val_loss: 1.6962 - val_accuracy: 0.3000
Epoch 111/200
9750 - val loss: 1.9599 - val accuracy: 0.3500
Epoch 112/200
2/2 [=============== ] - 0s 167ms/step - loss: 0.0123 - accuracy: 1.
0000 - val_loss: 1.7990 - val_accuracy: 0.2500
Epoch 113/200
2/2 [=============== ] - 0s 169ms/step - loss: 0.0109 - accuracy: 1.
0000 - val_loss: 1.7656 - val_accuracy: 0.3000
Epoch 114/200
2/2 [=============== ] - 0s 165ms/step - loss: 0.0193 - accuracy: 1.
0000 - val_loss: 1.7173 - val_accuracy: 0.3000
Epoch 115/200
2/2 [=============== ] - 0s 378ms/step - loss: 0.0679 - accuracy: 0.
9750 - val_loss: 1.8924 - val_accuracy: 0.4000
Epoch 116/200
2/2 [================ ] - 0s 166ms/step - loss: 0.0096 - accuracy: 1.
0000 - val_loss: 1.7928 - val_accuracy: 0.4500
Epoch 117/200
0000 - val loss: 2.0728 - val accuracy: 0.4500
Epoch 118/200
2/2 [=============== ] - 0s 346ms/step - loss: 0.0233 - accuracy: 1.
0000 - val_loss: 1.6017 - val_accuracy: 0.4500
Epoch 119/200
2/2 [================ ] - 0s 168ms/step - loss: 0.0171 - accuracy: 1.
0000 - val_loss: 1.2964 - val_accuracy: 0.5000
Epoch 120/200
2/2 [============ ] - 0s 163ms/step - loss: 0.0190 - accuracy: 1.
0000 - val_loss: 1.8980 - val_accuracy: 0.4000
Epoch 121/200
2/2 [============= ] - 0s 347ms/step - loss: 0.1152 - accuracy: 0.
9250 - val_loss: 1.3729 - val_accuracy: 0.4500
Epoch 122/200
0000 - val_loss: 1.5913 - val_accuracy: 0.4000
Epoch 123/200
0000 - val loss: 1.5685 - val accuracy: 0.4000
Epoch 124/200
2/2 [===============] - 0s 361ms/step - loss: 0.0226 - accuracy: 1.
0000 - val_loss: 1.4641 - val_accuracy: 0.4000
Epoch 125/200
2/2 [=================== ] - 0s 171ms/step - loss: 0.0114 - accuracy: 1.
0000 - val_loss: 1.3671 - val_accuracy: 0.4500
Epoch 126/200
0000 - val loss: 1.4092 - val accuracy: 0.4500
Epoch 127/200
0000 - val loss: 1.4888 - val accuracy: 0.4000
Epoch 128/200
2/2 [=================== ] - 0s 170ms/step - loss: 0.0111 - accuracy: 1.
0000 - val_loss: 1.4459 - val_accuracy: 0.4500
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Epoch 129/200
2/2 [============== ] - 0s 174ms/step - loss: 0.0215 - accuracy: 1.
0000 - val loss: 1.4732 - val accuracy: 0.4000
Epoch 130/200
2/2 [==============] - 0s 160ms/step - loss: 0.0139 - accuracy: 1.
0000 - val_loss: 1.4173 - val_accuracy: 0.4000
Epoch 131/200
0000 - val_loss: 1.3274 - val_accuracy: 0.5000
Epoch 132/200
2/2 [================ ] - 0s 167ms/step - loss: 0.0177 - accuracy: 1.
0000 - val_loss: 1.4717 - val_accuracy: 0.5000
Epoch 133/200
2/2 [============ ] - 0s 169ms/step - loss: 0.0263 - accuracy: 1.
0000 - val_loss: 1.3440 - val_accuracy: 0.4000
Epoch 134/200
2/2 [==============] - 0s 354ms/step - loss: 0.0077 - accuracy: 1.
0000 - val_loss: 1.4001 - val_accuracy: 0.4000
Epoch 135/200
2/2 [=============== ] - 0s 389ms/step - loss: 0.0276 - accuracy: 1.
0000 - val_loss: 1.4951 - val_accuracy: 0.4500
Epoch 136/200
2/2 [=============== ] - 0s 177ms/step - loss: 0.0063 - accuracy: 1.
0000 - val_loss: 1.4520 - val_accuracy: 0.4500
Epoch 137/200
2/2 [================ ] - 0s 166ms/step - loss: 0.0039 - accuracy: 1.
0000 - val_loss: 1.4438 - val_accuracy: 0.4500
Epoch 138/200
2/2 [============== ] - 0s 369ms/step - loss: 0.0161 - accuracy: 1.
0000 - val_loss: 1.2920 - val_accuracy: 0.4500
Epoch 139/200
2/2 [============= ] - 0s 166ms/step - loss: 0.0181 - accuracy: 1.
0000 - val_loss: 1.2206 - val_accuracy: 0.4500
Epoch 140/200
2/2 [==============] - 0s 167ms/step - loss: 0.0752 - accuracy: 0.
9750 - val_loss: 2.1110 - val_accuracy: 0.4000
Epoch 141/200
2/2 [=============== ] - 0s 362ms/step - loss: 0.0210 - accuracy: 1.
0000 - val_loss: 2.0432 - val_accuracy: 0.5500
Epoch 142/200
2/2 [=============== ] - 0s 347ms/step - loss: 0.0105 - accuracy: 1.
0000 - val_loss: 1.7050 - val_accuracy: 0.4000
Epoch 143/200
2/2 [================ ] - 0s 388ms/step - loss: 0.0159 - accuracy: 1.
0000 - val_loss: 1.7426 - val_accuracy: 0.4000
Epoch 144/200
2/2 [==============] - 0s 338ms/step - loss: 0.0059 - accuracy: 1.
0000 - val loss: 1.7546 - val accuracy: 0.4000
Epoch 145/200
2/2 [============= ] - 0s 170ms/step - loss: 0.0074 - accuracy: 1.
0000 - val loss: 1.8944 - val accuracy: 0.3500
Epoch 146/200
2/2 [================ ] - 0s 161ms/step - loss: 0.0162 - accuracy: 1.
0000 - val_loss: 1.9694 - val_accuracy: 0.4000
Epoch 147/200
0000 - val_loss: 1.9586 - val_accuracy: 0.4500
Epoch 148/200
2/2 [============= ] - 0s 173ms/step - loss: 0.0157 - accuracy: 1.
0000 - val loss: 1.9719 - val accuracy: 0.4000
Epoch 149/200
2/2 [=============] - 0s 347ms/step - loss: 0.0103 - accuracy: 1.
0000 - val_loss: 2.0627 - val_accuracy: 0.4000
Epoch 150/200
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0000 - val_loss: 2.0335 - val_accuracy: 0.4000
Epoch 151/200
0000 - val loss: 2.0520 - val accuracy: 0.4500
Epoch 152/200
2/2 [==============] - 0s 350ms/step - loss: 0.0348 - accuracy: 1.
0000 - val_loss: 1.7758 - val_accuracy: 0.4000
Epoch 153/200
2/2 [=============== ] - 0s 160ms/step - loss: 0.0094 - accuracy: 1.
0000 - val_loss: 1.7052 - val_accuracy: 0.4000
Epoch 154/200
2/2 [============ ] - 0s 169ms/step - loss: 0.0051 - accuracy: 1.
0000 - val loss: 1.7356 - val accuracy: 0.4000
Epoch 155/200
2/2 [==============] - 0s 168ms/step - loss: 0.0112 - accuracy: 1.
0000 - val_loss: 1.9303 - val_accuracy: 0.3500
Epoch 156/200
2/2 [=============== ] - 0s 163ms/step - loss: 0.0124 - accuracy: 1.
0000 - val_loss: 1.8441 - val_accuracy: 0.3500
Epoch 157/200
2/2 [============= ] - 0s 355ms/step - loss: 0.0064 - accuracy: 1.
0000 - val_loss: 1.9285 - val_accuracy: 0.4500
Epoch 158/200
2/2 [==============] - 0s 341ms/step - loss: 0.0102 - accuracy: 1.
0000 - val_loss: 1.7157 - val_accuracy: 0.4000
Epoch 159/200
2/2 [================ ] - 1s 611ms/step - loss: 0.0061 - accuracy: 1.
0000 - val_loss: 1.7069 - val_accuracy: 0.4500
Epoch 160/200
0000 - val_loss: 1.8899 - val_accuracy: 0.4500
Epoch 161/200
2/2 [==============] - 1s 166ms/step - loss: 0.0065 - accuracy: 1.
0000 - val_loss: 1.5438 - val_accuracy: 0.3000
Epoch 162/200
2/2 [================ ] - 1s 631ms/step - loss: 0.0070 - accuracy: 1.
0000 - val_loss: 1.5461 - val_accuracy: 0.3500
Epoch 163/200
2/2 [==============] - 0s 170ms/step - loss: 0.0067 - accuracy: 1.
0000 - val_loss: 1.7965 - val_accuracy: 0.4000
Epoch 164/200
2/2 [================ ] - 0s 352ms/step - loss: 0.0105 - accuracy: 1.
0000 - val_loss: 1.6793 - val_accuracy: 0.3500
Epoch 165/200
2/2 [================= ] - 0s 357ms/step - loss: 0.0148 - accuracy: 1.
0000 - val_loss: 1.6976 - val_accuracy: 0.4000
Epoch 166/200
2/2 [============= ] - 0s 341ms/step - loss: 0.0119 - accuracy: 1.
0000 - val loss: 2.9507 - val accuracy: 0.4500
2/2 [================ ] - 0s 172ms/step - loss: 0.0385 - accuracy: 1.
0000 - val_loss: 4.1275 - val_accuracy: 0.3000
Epoch 168/200
2/2 [============] - 0s 368ms/step - loss: 0.0362 - accuracy: 0.
9750 - val_loss: 2.2239 - val_accuracy: 0.4000
Epoch 169/200
2/2 [============= ] - 0s 161ms/step - loss: 0.0300 - accuracy: 0.
9750 - val loss: 1.1546 - val accuracy: 0.5000
Epoch 170/200
2/2 [================ ] - 0s 353ms/step - loss: 0.0061 - accuracy: 1.
0000 - val_loss: 1.1347 - val_accuracy: 0.5500
Epoch 171/200
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0000 - val_loss: 0.9618 - val_accuracy: 0.4500
Epoch 172/200
2/2 [============= ] - 0s 164ms/step - loss: 0.0035 - accuracy: 1.
0000 - val_loss: 0.9572 - val_accuracy: 0.4500
Epoch 173/200
2/2 [============ ] - 0s 349ms/step - loss: 0.0183 - accuracy: 1.
0000 - val_loss: 0.9145 - val_accuracy: 0.4500
Epoch 174/200
2/2 [==============] - 0s 348ms/step - loss: 0.0028 - accuracy: 1.
0000 - val_loss: 0.9013 - val_accuracy: 0.4500
Epoch 175/200
2/2 [==============] - 0s 163ms/step - loss: 0.0379 - accuracy: 0.
9750 - val loss: 1.0403 - val accuracy: 0.4500
Epoch 176/200
2/2 [=============== ] - 0s 353ms/step - loss: 0.0090 - accuracy: 1.
0000 - val_loss: 1.0402 - val_accuracy: 0.4500
Epoch 177/200
2/2 [=============== ] - 0s 171ms/step - loss: 0.0054 - accuracy: 1.
0000 - val_loss: 1.0385 - val_accuracy: 0.4500
Epoch 178/200
2/2 [=============== ] - 0s 351ms/step - loss: 0.0469 - accuracy: 0.
9750 - val_loss: 1.0831 - val_accuracy: 0.5500
Epoch 179/200
2/2 [=============== ] - 0s 164ms/step - loss: 0.0243 - accuracy: 0.
9750 - val_loss: 1.0463 - val_accuracy: 0.4500
Epoch 180/200
2/2 [================ ] - 0s 165ms/step - loss: 0.0080 - accuracy: 1.
0000 - val_loss: 1.0216 - val_accuracy: 0.4500
Epoch 181/200
0000 - val loss: 1.0089 - val accuracy: 0.5000
Epoch 182/200
2/2 [=============== ] - 0s 368ms/step - loss: 0.0063 - accuracy: 1.
0000 - val_loss: 0.9628 - val_accuracy: 0.5000
Epoch 183/200
2/2 [================ ] - 0s 356ms/step - loss: 0.0287 - accuracy: 0.
9750 - val_loss: 0.9455 - val_accuracy: 0.5000
Epoch 184/200
2/2 [============= ] - 0s 355ms/step - loss: 0.0989 - accuracy: 0.
9250 - val_loss: 0.9614 - val_accuracy: 0.5000
Epoch 185/200
2/2 [============= ] - 0s 352ms/step - loss: 0.0139 - accuracy: 1.
0000 - val_loss: 0.9310 - val_accuracy: 0.5000
Epoch 186/200
0000 - val_loss: 0.9062 - val_accuracy: 0.5000
Epoch 187/200
2/2 [============= ] - 0s 366ms/step - loss: 0.0063 - accuracy: 1.
0000 - val loss: 0.8981 - val accuracy: 0.5000
Epoch 188/200
2/2 [==============] - 0s 169ms/step - loss: 0.0086 - accuracy: 1.
0000 - val_loss: 0.9006 - val_accuracy: 0.5000
Epoch 189/200
2/2 [=================== ] - 0s 180ms/step - loss: 0.0136 - accuracy: 1.
0000 - val_loss: 0.9073 - val_accuracy: 0.5000
Epoch 190/200
2/2 [============= ] - 0s 350ms/step - loss: 0.0080 - accuracy: 1.
0000 - val loss: 1.0335 - val accuracy: 0.4000
Epoch 191/200
2/2 [==============] - 0s 360ms/step - loss: 0.0042 - accuracy: 1.
0000 - val loss: 0.9818 - val accuracy: 0.4000
Epoch 192/200
2/2 [=================== ] - 0s 358ms/step - loss: 0.0356 - accuracy: 0.
9750 - val_loss: 0.8645 - val_accuracy: 0.5000
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Epoch 193/200
       2/2 [============== ] - 0s 349ms/step - loss: 0.0105 - accuracy: 1.
       0000 - val loss: 0.8613 - val accuracy: 0.5000
       Epoch 194/200
       9750 - val loss: 0.8604 - val accuracy: 0.5500
       Epoch 195/200
       0000 - val_loss: 0.8565 - val_accuracy: 0.5500
       Epoch 196/200
       2/2 [================ ] - 0s 168ms/step - loss: 0.0044 - accuracy: 1.
       0000 - val_loss: 0.8863 - val_accuracy: 0.5000
       Epoch 197/200
       2/2 [============= ] - 0s 368ms/step - loss: 0.0142 - accuracy: 1.
       0000 - val_loss: 1.0028 - val_accuracy: 0.5000
       Epoch 198/200
       2/2 [================ ] - 0s 168ms/step - loss: 0.0140 - accuracy: 1.
       0000 - val_loss: 1.0251 - val_accuracy: 0.5000
       Epoch 199/200
       2/2 [================ ] - 0s 173ms/step - loss: 0.0210 - accuracy: 1.
       0000 - val_loss: 0.9973 - val_accuracy: 0.4500
       Epoch 200/200
       2/2 [================ ] - 0s 355ms/step - loss: 0.1251 - accuracy: 0.
       9500 - val_loss: 0.9060 - val_accuracy: 0.5500
In [12]: loss, accuracy = model.evaluate(test_images)
       print("Accuracy of the model:", accuracy)
       1/1 [=============== ] - 0s 137ms/step - loss: 0.9060 - accuracy: 0.
       5500
       Accuracy of the model: 0.550000011920929
In [11]:
       from sklearn.metrics import confusion_matrix
       import seaborn as sns
       preds = model.predict(test_images)
       pred_labels = np.argmax(preds, axis=1)
       print("Actual labels: ", test_images.labels)
       print("Predicted labels: ", pred_labels)
       Actual labels: [0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1]
```