```
from tensorflow.keras.layers import Dense, Flatten, Input
from tensorflow.keras.models import Model
from tensorflow.keras.preprocessing import image
from keras.preprocessing.image import ImageDataGenerator,load imageDataGenerat
from keras.applications.vgq16 import VGG16, preprocess input
from glob import glob
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras import layers, models
WARNING:tensorflow:From c:\Users\heman\AppData\Local\Programs\Python\
Python311\Lib\site-packages\keras\src\losses.py:2976: The name
tf.losses.sparse softmax cross entropy is deprecated. Please use
tf.compat.v1.losses.sparse softmax cross entropy instead.
vgg=VGG16(weights="imagenet",include top=False,input shape=(224, 224,
3))
2023-11-25 15:53:39.194393: W
tensorflow/core/common runtime/gpu/gpu device.cc:2256] Cannot dlopen
some GPU libraries. Please make sure the missing libraries mentioned
above are installed properly if you would like to use GPU. Follow the
guide at https://www.tensorflow.org/install/gpu for how to download
and setup the required libraries for your platform.
Skipping registering GPU devices...
for layer in vgg.layers:
         layer.trainable=False
# x=Flatten()(vgg.output)
model = models.Sequential()
model.add(vqq)
# model.add
model.add(layers.Flatten())
model.add(layers.Dense(256, activation='relu'))
model.add(layers.Dense(100, activation='softmax'))
model.compile(
        loss="categorical crossentropy",
        optimizer="adam",
        metrics=["accuracy"],run eagerly=True
)
from tensorflow.keras.preprocessing.image import ImageDataGenerator
train datagen=ImageDataGenerator(rescale = 1./255,
                                                                         shear range =0.2,
                                                                         zoom range = 0.2,
                                                                         horizontal flip = True)
```

```
test datagen = ImageDataGenerator(rescale = 1./255)
# val datagen=ImageDataGenerator(rescale = 1./255,
                              shear range =0.2,
#
                              zoom\ range = 0.2,
                              horizontal flip = True)
training set=train datagen.flow from directory(r"/home/tihan/Desktop/
hemanth/archive/
train/",target_size=(224,224),batch size=10,shuffle=True,class mode="c
ategorical")
test set=test datagen.flow from directory(r"/home/tihan/Desktop/hemant
h/archive/
test/",target_size=(224,224),batch_size=10,class mode="categorical")
# valid set = val datagen.flow from directory(
     directory=r"/home/tihan/Desktop/hemanth/archive/valid/",
#
     target size=(224, 224),
#
     batch size=10,
#
     class mode="categorical",
#
     shuffle=True,
# )
Found 13492 images belonging to 100 classes.
Found 500 images belonging to 100 classes.
import sys
# fit the mdel
r=model.fit generator(training set,
validation data=test set,
epochs=200,
steps per epoch=len(training set)//3,
validation steps=len(test set)//3
Epoch 1/200
/tmp/ipykernel 21375/2596593733.py:3: UserWarning:
`Model.fit generator` is deprecated and will be removed in a future
version. Please use `Model.fit`, which supports generators.
 r=model.fit generator(training set,
4.2676 - accuracy: 0.0591 - val loss: 3.6382 - val accuracy: 0.0875
Epoch 2/200
3.2785 - accuracy: 0.1822 - val loss: 2.6735 - val accuracy: 0.2625
Epoch 3/200
2.7061 - accuracy: 0.2832 - val loss: 2.2210 - val accuracy: 0.3063
Epoch 4/200
```

```
2.4545 - accuracy: 0.3253 - val_loss: 2.1669 - val_accuracy: 0.4250
Epoch 5/200
2.2925 - accuracy: 0.3740 - val loss: 1.9237 - val accuracy: 0.4938
Epoch 6/200
2.1676 - accuracy: 0.3927 - val loss: 1.8678 - val accuracy: 0.4688
Epoch 7/200
2.0998 - accuracy: 0.4067 - val loss: 1.8781 - val accuracy: 0.4437
Epoch 8/200
1.9946 - accuracy: 0.4402 - val loss: 1.9555 - val accuracy: 0.4750
Epoch 9/200
1.8953 - accuracy: 0.4378 - val loss: 1.8748 - val accuracy: 0.4563
Epoch 10/200
1.8659 - accuracy: 0.4607 - val loss: 1.7351 - val accuracy: 0.4812
Epoch 11/200
1.8378 - accuracy: 0.4735 - val loss: 1.7104 - val accuracy: 0.4938
Epoch 12/200
1.7520 - accuracy: 0.4918 - val loss: 1.6310 - val accuracy: 0.5312
Epoch 13/200
1.7618 - accuracy: 0.4929 - val_loss: 1.6747 - val_accuracy: 0.4688
Epoch 14/200
- accuracy: 0.5087 - val loss: 1.7295 - val_accuracy: 0.4688
Epoch 15/200
1.6510 - accuracy: 0.5118 - val loss: 1.6426 - val accuracy: 0.4938
Epoch 16/200
1.6200 - accuracy: 0.5200 - val loss: 1.6795 - val accuracy: 0.4812
Epoch 17/200
1.6064 - accuracy: 0.5169 - val loss: 1.7931 - val accuracy: 0.5125
Epoch 18/200
1.5628 - accuracy: 0.5452 - val_loss: 1.5617 - val_accuracy: 0.4938
Epoch 19/200
- accuracy: 0.5387 - val_loss: 1.8110 - val_accuracy: 0.4688
Epoch 20/200
1.5755 - accuracy: 0.5376 - val loss: 1.8329 - val accuracy: 0.4688
```

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Epoch 21/200
1.5092 - accuracy: 0.5528 - val loss: 1.6329 - val accuracy: 0.5375
Epoch 22/200
1.4891 - accuracy: 0.5569 - val loss: 1.5821 - val accuracy: 0.5750
Epoch 23/200
1.4685 - accuracy: 0.5536 - val loss: 1.6719 - val accuracy: 0.5250
Epoch 24/200
1.4457 - accuracy: 0.5764 - val loss: 1.6311 - val accuracy: 0.5250
Epoch 25/200
1.4168 - accuracy: 0.5749 - val_loss: 1.7612 - val_accuracy: 0.5188
Epoch 26/200
1.4085 - accuracy: 0.5731 - val loss: 1.6913 - val accuracy: 0.5562
Epoch 27/200
1.4027 - accuracy: 0.5730 - val loss: 1.8899 - val accuracy: 0.5312
Epoch 28/200
1.4064 - accuracy: 0.5808 - val loss: 1.4493 - val accuracy: 0.5375
Epoch 29/200
1.3695 - accuracy: 0.5869 - val_loss: 1.7067 - val_accuracy: 0.5562
Epoch 30/200
1.3462 - accuracy: 0.5882 - val loss: 1.6785 - val accuracy: 0.5625
Epoch 31/200
1.3327 - accuracy: 0.6087 - val loss: 1.4835 - val accuracy: 0.5375
Epoch 32/200
1.3021 - accuracy: 0.6069 - val loss: 1.6307 - val accuracy: 0.5688
Epoch 33/200
1.3019 - accuracy: 0.6020 - val_loss: 1.5952 - val_accuracy: 0.5500
Epoch 34/200
1.3199 - accuracy: 0.6040 - val loss: 1.8691 - val accuracy: 0.4938
Epoch 35/200
1.2809 - accuracy: 0.6187 - val loss: 1.4842 - val accuracy: 0.5750
Epoch 36/200
1.2732 - accuracy: 0.6084 - val loss: 1.3189 - val accuracy: 0.5813
Epoch 37/200
```

```
1.3066 - accuracy: 0.6042 - val loss: 1.6952 - val accuracy: 0.5938
Epoch 38/200
1.2204 - accuracy: 0.6262 - val_loss: 1.5685 - val_accuracy: 0.5375
Epoch 39/200
1.2159 - accuracy: 0.6391 - val loss: 2.1266 - val accuracy: 0.4625
Epoch 40/200
1.1925 - accuracy: 0.6416 - val loss: 1.8459 - val accuracy: 0.4938
Epoch 41/200
1.2578 - accuracy: 0.6189 - val loss: 1.6531 - val accuracy: 0.5625
Epoch 42/200
1.2284 - accuracy: 0.6218 - val loss: 1.7149 - val accuracy: 0.5625
Epoch 43/200
1.1561 - accuracy: 0.6376 - val loss: 1.3801 - val accuracy: 0.6062
Epoch 44/200
1.1926 - accuracy: 0.6369 - val loss: 1.5008 - val accuracy: 0.6250
Epoch 45/200
1.1962 - accuracy: 0.6356 - val loss: 1.7175 - val accuracy: 0.5125
Epoch 46/200
1.1418 - accuracy: 0.6512 - val loss: 1.8064 - val accuracy: 0.4875
Epoch 47/200
1.2331 - accuracy: 0.6322 - val loss: 1.6999 - val accuracy: 0.5562
Epoch 48/200
1.1844 - accuracy: 0.6371 - val loss: 1.6218 - val accuracy: 0.5750
Epoch 49/200
1.1040 - accuracy: 0.6598 - val loss: 1.5878 - val accuracy: 0.5875
Epoch 50/200
1.2055 - accuracy: 0.6329 - val loss: 1.6383 - val accuracy: 0.5562
Epoch 51/200
1.0758 - accuracy: 0.6691 - val loss: 1.9525 - val accuracy: 0.5437
Epoch 52/200
1.1245 - accuracy: 0.6578 - val loss: 1.7306 - val accuracy: 0.5500
Epoch 53/200
```

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1.1119 - accuracy: 0.6600 - val loss: 1.4806 - val accuracy: 0.5688
Epoch 54/200
1.1263 - accuracy: 0.6533 - val loss: 1.5374 - val accuracy: 0.5875
Epoch 55/200
1.1167 - accuracy: 0.6592 - val loss: 1.7258 - val accuracy: 0.5938
Epoch 56/200
1.1193 - accuracy: 0.6602 - val loss: 1.7258 - val accuracy: 0.4750
Epoch 57/200
1.0639 - accuracy: 0.6802 - val loss: 1.8347 - val accuracy: 0.5312
Epoch 58/200
1.0880 - accuracy: 0.6609 - val loss: 1.5999 - val accuracy: 0.5437
Epoch 59/200
1.0641 - accuracy: 0.6684 - val loss: 1.9036 - val accuracy: 0.5875
Epoch 60/200
1.0832 - accuracy: 0.6673 - val loss: 1.8469 - val accuracy: 0.5562
Epoch 61/200
1.0805 - accuracy: 0.6691 - val loss: 1.8139 - val_accuracy: 0.5312
Epoch 62/200
1.0718 - accuracy: 0.6672 - val loss: 1.8753 - val accuracy: 0.5250
Epoch 63/200
1.0302 - accuracy: 0.6773 - val loss: 1.6175 - val accuracy: 0.5875
Epoch 64/200
1.0705 - accuracy: 0.6719 - val loss: 1.7144 - val accuracy: 0.5625
Epoch 65/200
1.0408 - accuracy: 0.6800 - val loss: 1.5280 - val accuracy: 0.6125
Epoch 66/200
1.0393 - accuracy: 0.6812 - val loss: 1.8137 - val accuracy: 0.5437
Epoch 67/200
1.0379 - accuracy: 0.6740 - val loss: 1.8723 - val accuracy: 0.5250
Epoch 68/200
1.0789 - accuracy: 0.6692 - val_loss: 1.9918 - val_accuracy: 0.5000
Epoch 69/200
1.0057 - accuracy: 0.6853 - val loss: 1.6071 - val accuracy: 0.5875
```

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Epoch 70/200
0.9980 - accuracy: 0.6858 - val loss: 1.4434 - val accuracy: 0.5562
Epoch 71/200
0.9931 - accuracy: 0.6903 - val loss: 1.8019 - val accuracy: 0.5250
Epoch 72/200
1.0229 - accuracy: 0.6871 - val loss: 1.9035 - val accuracy: 0.5625
Epoch 73/200
1.0715 - accuracy: 0.6671 - val loss: 2.0968 - val accuracy: 0.5063
Epoch 74/200
1.0121 - accuracy: 0.6848 - val_loss: 1.8226 - val_accuracy: 0.5312
Epoch 75/200
1.0093 - accuracy: 0.6759 - val loss: 2.3606 - val accuracy: 0.4062
Epoch 76/200
0.9445 - accuracy: 0.7044 - val loss: 1.7714 - val accuracy: 0.5312
Epoch 77/200
0.9525 - accuracy: 0.7046 - val loss: 2.0103 - val accuracy: 0.5188
Epoch 78/200
1.0001 - accuracy: 0.6929 - val_loss: 1.7192 - val_accuracy: 0.6187
Epoch 79/200
1.0065 - accuracy: 0.6827 - val loss: 1.8685 - val accuracy: 0.5938
Epoch 80/200
0.9656 - accuracy: 0.6979 - val loss: 1.4279 - val accuracy: 0.5750
Epoch 81/200
0.9863 - accuracy: 0.6929 - val loss: 2.0428 - val accuracy: 0.5625
Epoch 82/200
0.9740 - accuracy: 0.6946 - val loss: 2.1676 - val accuracy: 0.5437
Epoch 83/200
0.9736 - accuracy: 0.6886 - val loss: 1.8462 - val accuracy: 0.5437
Epoch 84/200
0.9579 - accuracy: 0.7082 - val loss: 1.9613 - val accuracy: 0.5437
Epoch 85/200
0.9732 - accuracy: 0.7001 - val loss: 2.2883 - val accuracy: 0.4812
Epoch 86/200
```

```
0.9761 - accuracy: 0.6932 - val loss: 1.9391 - val accuracy: 0.5875
Epoch 87/200
0.9640 - accuracy: 0.7022 - val_loss: 1.5999 - val_accuracy: 0.6125
Epoch 88/200
0.9399 - accuracy: 0.7087 - val loss: 2.1144 - val accuracy: 0.4938
Epoch 89/200
0.9249 - accuracy: 0.7087 - val loss: 1.9723 - val accuracy: 0.5500
Epoch 90/200
0.9337 - accuracy: 0.7178 - val loss: 1.5982 - val accuracy: 0.5688
Epoch 91/200
0.9429 - accuracy: 0.7069 - val loss: 2.0088 - val accuracy: 0.5250
Epoch 92/200
0.9469 - accuracy: 0.6991 - val loss: 2.2691 - val accuracy: 0.5063
Epoch 93/200
0.9538 - accuracy: 0.6997 - val loss: 1.7305 - val accuracy: 0.5500
Epoch 94/200
0.9154 - accuracy: 0.7084 - val loss: 1.7469 - val accuracy: 0.5625
Epoch 95/200
0.9229 - accuracy: 0.7064 - val loss: 1.9126 - val accuracy: 0.5250
Epoch 96/200
0.9156 - accuracy: 0.7171 - val loss: 1.9309 - val accuracy: 0.5813
Epoch 97/200
0.9107 - accuracy: 0.7173 - val loss: 1.7972 - val accuracy: 0.5000
Epoch 98/200
0.9163 - accuracy: 0.7120 - val loss: 2.0139 - val accuracy: 0.5500
Epoch 99/200
0.8535 - accuracy: 0.7275 - val loss: 2.2656 - val accuracy: 0.5125
Epoch 100/200
0.8714 - accuracy: 0.7264 - val loss: 2.0573 - val accuracy: 0.5375
Epoch 101/200
0.9073 - accuracy: 0.7197 - val loss: 1.5998 - val accuracy: 0.5500
Epoch 102/200
```

```
0.9121 - accuracy: 0.7095 - val loss: 2.2167 - val accuracy: 0.5813
Epoch 103/200
0.8696 - accuracy: 0.7173 - val_loss: 1.8655 - val accuracy: 0.5938
Epoch 104/200
0.9181 - accuracy: 0.7131 - val loss: 2.1975 - val accuracy: 0.5625
Epoch 105/200
0.9115 - accuracy: 0.7150 - val loss: 1.9594 - val accuracy: 0.5750
Epoch 106/200
0.9150 - accuracy: 0.7064 - val loss: 2.1357 - val accuracy: 0.5312
Epoch 107/200
0.8889 - accuracy: 0.7242 - val loss: 1.8028 - val accuracy: 0.5188
Epoch 108/200
0.9078 - accuracy: 0.7220 - val loss: 1.7124 - val accuracy: 0.5938
Epoch 109/200
0.8633 - accuracy: 0.7222 - val loss: 2.0086 - val accuracy: 0.5312
Epoch 110/200
0.8704 - accuracy: 0.7224 - val loss: 1.9382 - val accuracy: 0.5375
Epoch 111/200
0.8876 - accuracy: 0.7180 - val loss: 2.2552 - val accuracy: 0.5312
Epoch 112/200
0.9183 - accuracy: 0.7027 - val loss: 1.9243 - val accuracy: 0.5813
Epoch 113/200
0.8474 - accuracy: 0.7409 - val loss: 1.9919 - val accuracy: 0.5625
Epoch 114/200
0.8554 - accuracy: 0.7373 - val loss: 1.7321 - val accuracy: 0.5938
Epoch 115/200
0.8757 - accuracy: 0.7198 - val loss: 2.3421 - val accuracy: 0.5813
Epoch 116/200
0.8611 - accuracy: 0.7324 - val loss: 1.8934 - val accuracy: 0.6062
Epoch 117/200
0.8399 - accuracy: 0.7331 - val_loss: 1.8726 - val_accuracy: 0.5875
Epoch 118/200
0.8810 - accuracy: 0.7287 - val loss: 2.0049 - val accuracy: 0.5562
```

```
Epoch 119/200
0.8692 - accuracy: 0.7317 - val loss: 1.4585 - val accuracy: 0.6438
Epoch 120/200
0.8260 - accuracy: 0.7320 - val loss: 2.0452 - val accuracy: 0.5500
Epoch 121/200
0.8732 - accuracy: 0.7237 - val loss: 2.3288 - val accuracy: 0.5813
Epoch 122/200
0.8592 - accuracy: 0.7320 - val loss: 1.9264 - val accuracy: 0.5625
Epoch 123/200
0.8476 - accuracy: 0.7346 - val_loss: 1.9989 - val_accuracy: 0.5312
Epoch 124/200
0.9137 - accuracy: 0.7248 - val loss: 2.0361 - val accuracy: 0.5813
Epoch 125/200
0.8092 - accuracy: 0.7438 - val loss: 2.1395 - val accuracy: 0.5750
Epoch 126/200
0.8508 - accuracy: 0.7329 - val loss: 1.9098 - val accuracy: 0.5437
Epoch 127/200
0.8431 - accuracy: 0.7396 - val_loss: 1.6813 - val_accuracy: 0.6250
Epoch 128/200
0.8297 - accuracy: 0.7420 - val loss: 2.2508 - val accuracy: 0.4875
Epoch 129/200
0.8524 - accuracy: 0.7298 - val_loss: 2.2537 - val_accuracy: 0.5625
Epoch 130/200
0.8306 - accuracy: 0.7349 - val loss: 2.1398 - val accuracy: 0.5437
Epoch 131/200
0.8320 - accuracy: 0.7349 - val loss: 2.0287 - val accuracy: 0.6000
Epoch 132/200
0.8167 - accuracy: 0.7400 - val loss: 1.9116 - val accuracy: 0.6125
Epoch 133/200
0.8310 - accuracy: 0.7304 - val loss: 2.0843 - val accuracy: 0.6125
Epoch 134/200
0.8422 - accuracy: 0.7358 - val loss: 2.4390 - val accuracy: 0.5125
Epoch 135/200
```

```
0.8120 - accuracy: 0.7542 - val loss: 2.2188 - val accuracy: 0.5250
Epoch 136/200
0.8120 - accuracy: 0.7507 - val loss: 1.9734 - val accuracy: 0.5688
Epoch 137/200
0.8084 - accuracy: 0.7471 - val loss: 1.8624 - val accuracy: 0.6250
Epoch 138/200
0.8181 - accuracy: 0.7449 - val loss: 1.8841 - val accuracy: 0.5813
Epoch 139/200
0.8412 - accuracy: 0.7369 - val loss: 2.1582 - val accuracy: 0.5312
Epoch 140/200
0.8295 - accuracy: 0.7453 - val loss: 1.9972 - val accuracy: 0.5562
Epoch 141/200
0.8022 - accuracy: 0.7487 - val loss: 1.7588 - val accuracy: 0.5688
Epoch 142/200
0.7810 - accuracy: 0.7540 - val loss: 2.2691 - val accuracy: 0.6000
Epoch 143/200
0.8121 - accuracy: 0.7469 - val loss: 1.9227 - val accuracy: 0.5813
Epoch 144/200
0.7797 - accuracy: 0.7493 - val loss: 2.1357 - val accuracy: 0.5188
Epoch 145/200
0.8582 - accuracy: 0.7344 - val loss: 1.7512 - val accuracy: 0.6438
Epoch 146/200
0.7995 - accuracy: 0.7562 - val loss: 1.9515 - val accuracy: 0.5688
Epoch 147/200
0.7953 - accuracy: 0.7518 - val loss: 2.0580 - val accuracy: 0.5562
Epoch 148/200
0.8006 - accuracy: 0.7444 - val loss: 2.1643 - val accuracy: 0.5375
Epoch 149/200
0.8021 - accuracy: 0.7429 - val loss: 2.6053 - val accuracy: 0.5437
Epoch 150/200
0.8254 - accuracy: 0.7440 - val loss: 1.8205 - val accuracy: 0.6000
Epoch 151/200
```

```
0.8014 - accuracy: 0.7460 - val loss: 1.9444 - val accuracy: 0.5688
Epoch 152/200
0.7900 - accuracy: 0.7644 - val_loss: 2.1073 - val accuracy: 0.5813
Epoch 153/200
0.7639 - accuracy: 0.7524 - val loss: 2.0046 - val accuracy: 0.5938
Epoch 154/200
0.7821 - accuracy: 0.7516 - val loss: 2.2472 - val accuracy: 0.5312
Epoch 155/200
0.7668 - accuracy: 0.7502 - val loss: 2.2663 - val accuracy: 0.5250
Epoch 156/200
0.7881 - accuracy: 0.7500 - val loss: 2.1956 - val accuracy: 0.5500
Epoch 157/200
0.8074 - accuracy: 0.7547 - val loss: 2.4630 - val accuracy: 0.5437
Epoch 158/200
0.7959 - accuracy: 0.7549 - val loss: 2.0413 - val accuracy: 0.5437
Epoch 159/200
0.7948 - accuracy: 0.7545 - val loss: 2.7265 - val accuracy: 0.4812
Epoch 160/200
0.7903 - accuracy: 0.7542 - val loss: 2.5283 - val accuracy: 0.5125
Epoch 161/200
0.8058 - accuracy: 0.7442 - val loss: 2.1723 - val_accuracy: 0.5375
Epoch 162/200
0.7452 - accuracy: 0.7593 - val loss: 2.2589 - val accuracy: 0.5125
Epoch 163/200
0.7563 - accuracy: 0.7687 - val loss: 1.5125 - val accuracy: 0.6375
Epoch 164/200
0.8114 - accuracy: 0.7500 - val loss: 2.2069 - val accuracy: 0.5875
Epoch 165/200
0.7719 - accuracy: 0.7580 - val loss: 2.2099 - val accuracy: 0.5688
Epoch 166/200
0.7546 - accuracy: 0.7571 - val_loss: 2.0882 - val_accuracy: 0.5875
Epoch 167/200
0.7763 - accuracy: 0.7596 - val loss: 2.2692 - val accuracy: 0.5125
Epoch 168/200
```

```
0.7438 - accuracy: 0.7656 - val loss: 2.3525 - val accuracy: 0.5562
Epoch 169/200
0.7844 - accuracy: 0.7427 - val loss: 2.3148 - val accuracy: 0.5625
Epoch 170/200
0.7564 - accuracy: 0.7624 - val loss: 1.9498 - val accuracy: 0.5875
Epoch 171/200
0.7649 - accuracy: 0.7565 - val loss: 1.8993 - val accuracy: 0.5938
Epoch 172/200
0.7632 - accuracy: 0.7567 - val loss: 2.5805 - val accuracy: 0.5375
Epoch 173/200
0.7525 - accuracy: 0.7600 - val loss: 2.5832 - val_accuracy: 0.5125
Epoch 174/200
0.7563 - accuracy: 0.7607 - val loss: 2.2898 - val accuracy: 0.5875
Epoch 175/200
0.7712 - accuracy: 0.7569 - val loss: 2.3997 - val accuracy: 0.5813
Epoch 176/200
0.7230 - accuracy: 0.7724 - val loss: 2.2547 - val accuracy: 0.5375
Epoch 177/200
0.7462 - accuracy: 0.7678 - val loss: 2.7413 - val accuracy: 0.5375
Epoch 178/200
0.7319 - accuracy: 0.7636 - val loss: 2.2247 - val accuracy: 0.5312
Epoch 179/200
0.7414 - accuracy: 0.7669 - val loss: 2.0316 - val accuracy: 0.5625
Epoch 180/200
0.7318 - accuracy: 0.7733 - val loss: 2.1588 - val accuracy: 0.6000
Epoch 181/200
0.7297 - accuracy: 0.7698 - val loss: 1.9641 - val accuracy: 0.5813
Epoch 182/200
0.7682 - accuracy: 0.7576 - val loss: 1.9650 - val accuracy: 0.6438
Epoch 183/200
0.7451 - accuracy: 0.7656 - val loss: 2.0734 - val accuracy: 0.6313
Epoch 184/200
```

```
0.7395 - accuracy: 0.7707 - val loss: 2.1681 - val accuracy: 0.5813
Epoch 185/200
0.7907 - accuracy: 0.7491 - val loss: 2.4809 - val accuracy: 0.5562
Epoch 186/200
0.7165 - accuracy: 0.7604 - val loss: 2.2217 - val accuracy: 0.5375
Epoch 187/200
0.7319 - accuracy: 0.7676 - val loss: 2.5507 - val accuracy: 0.5125
Epoch 188/200
0.7122 - accuracy: 0.7720 - val loss: 2.3414 - val accuracy: 0.5437
Epoch 189/200
0.7354 - accuracy: 0.7700 - val loss: 2.2924 - val accuracy: 0.5250
Epoch 190/200
0.7276 - accuracy: 0.7649 - val loss: 2.3497 - val accuracy: 0.5688
Epoch 191/200
0.7743 - accuracy: 0.7591 - val loss: 2.0234 - val accuracy: 0.5875
Epoch 192/200
0.7446 - accuracy: 0.7711 - val loss: 2.1069 - val accuracy: 0.5437
Epoch 193/200
0.7154 - accuracy: 0.7771 - val loss: 2.7973 - val accuracy: 0.5125
Epoch 194/200
0.7404 - accuracy: 0.7660 - val loss: 2.1837 - val accuracy: 0.5250
Epoch 195/200
0.7174 - accuracy: 0.7731 - val loss: 2.4318 - val accuracy: 0.5312
Epoch 196/200
0.7102 - accuracy: 0.7705 - val loss: 2.3180 - val accuracy: 0.6500
Epoch 197/200
0.7481 - accuracy: 0.7904 - val loss: 2.5570 - val accuracy: 0.7050
Epoch 198/200
0.7135 - accuracy: 0.7867 - val loss: 2.3970 - val accuracy: 0.7225
Epoch 199/200
0.7464 - accuracy: 0.8002 - val loss: 2.4535 - val accuracy: 0.7537
Epoch 200/200
0.6975 - accuracy: 0.8584 - val loss: 2.2365 - val accuracy: 0.8062
```

```
#Save Model
model.save("sports classification.h5")
/home/tihan/.local/lib/python3.10/site-packages/keras/src/engine/
training.py:3103: UserWarning: You are saving your model as an HDF5
file via `model.save()`. This file format is considered legacy. We
recommend using instead the native Keras format, e.g.
`model.save('my model.keras')`.
  saving api.save model(
from tensorflow.keras.models import load model
model=load_model("sports classification.h5")
WARNING:tensorflow:From c:\Users\heman\AppData\Local\Programs\Python\
Python311\Lib\site-packages\keras\src\backend.py:1398: The name
tf.executing eagerly outside functions is deprecated. Please use
tf.compat.v1.executing_eagerly_outside_functions instead.
WARNING:tensorflow:From c:\Users\heman\AppData\Local\Programs\Python\
Python311\Lib\site-packages\keras\src\layers\pooling\
max pooling2d.py:161: The name tf.nn.max pool is deprecated. Please
use tf.nn.max pool2d instead.
img=image.load img(r"D:\smartbridge data science\project\main\dataset\
test\boxing\2.jpg",target_size=(224,224))
x=image.img to array(img)
import numpy as np
x=np.expand dims(x,axis=0)
img data=preprocess input(x)
output=np.argmax(model.predict(img data),axis=1)
index=['air hockey', 'ampute football', 'archery', 'arm wrestling',
'axe throwing', 'balance beam', 'barell racing', 'baseball',
        'basketball', 'baton twirling', 'bike polo', 'billiards',
'bmx', 'bobsled', 'bowling', 'boxing', 'bull riding', 'bungee
jumping',
        'canoe slamon', 'cheerleading', 'chuckwagon racing', 'cricket',
'croquet', 'curling', 'disc golf', 'fencing', 'field hockey', 'figure
skating men',
'figure skating pairs', 'figure skating women', 'fly fishing', 'football', 'formula 1 racing', 'frisbee', 'gaga', 'giant slalom', 'golf', 'hammer throw', 'hang gliding', 'harness racing', 'high
jump', 'hockey', 'horse jumping', 'horse racing', 'horseshoe
pitching',
        'hurdles', 'hydroplane racing', 'ice climbing', 'ice yachting',
'jai alai', 'javelin', 'jousting', 'judo', 'lacrosse', 'log rolling', 'luge', 'motorcycle racing', 'mushing', 'nascar racing',
'olympic wrestling', 'parallel bar', 'pole climbing', 'pole dancing',
'pole vault',
        'polo', 'pommel horse', 'rings', 'rock climbing', 'roller
```

```
derby', 'rollerblade racing', 'rowing', 'rugby', 'sailboat racing',
'shot put',
       'shuffleboard', 'sidecar racing', 'ski jumping', 'sky surfing',
'skydiving', 'snow boarding', 'snowmobile racing', 'speed skating',
'steer wrestling',
      'sumo wrestling', 'surfing', 'swimming', 'table tennis',
'tennis', 'track bicycle', 'trapeze', 'tug of war', 'ultimate', 'uneven bars', 'volleyball',
       'water cycling', 'water polo', 'weightlifting', 'wheelchair
basketball', 'wheelchair racing', 'wingsuit flying']
result=str(index[output[0]])
result
'boxing'
```