MobileNet Model

Action Classes - 5

```
In [1]: from keras import models
   from keras.layers import Dense,Flatten
   from keras import backend as K
   import numpy as np
   import matplotlib.pyplot as plt

from keras.applications import mobilenet
```

```
In [2]: import tensorflow as tf
print("Num GPUs Available: ", len(tf.config.list_physical_devices('GPU'))
```

Num GPUs Available: 1

2022-08-25 22:35:43.901535: I tensorflow/stream_executor/cuda/cuda_gpu_ex ecutor.cc:975] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-08-25 22:35:43.930159: I tensorflow/stream_executor/cuda/cuda_gpu_ex ecutor.cc:975] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-08-25 22:35:43.930418: I tensorflow/stream_executor/cuda/cuda_gpu_ex ecutor.cc:975] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero

Dataset

```
In [3]: from keras.preprocessing.image import ImageDataGenerator

dataset_path = "./frames/"
    # will contain the categories in respective folders

# Data generators
train_datagen = ImageDataGenerator(rescale=1/255, validation_split=0.2)
```

```
In [4]: image size = (224,224)
        batch size = 10
        train batches = train datagen.flow from directory(
            dataset path,
            target_size = image_size,
            batch_size = batch_size,
            class mode = "categorical",
            subset = "training"
        )
        validation batches = train datagen.flow from directory(
            dataset path,
            target_size = image_size,
            batch size = batch size,
            class mode = "categorical",
            subset = "validation"
        test batches = train datagen.flow from directory(
            dataset path,
            target_size = image_size,
            batch size = batch size,
            class mode = "categorical",
            subset = "validation"
        Found 1546 images belonging to 5 classes.
        Found 384 images belonging to 5 classes.
        Found 384 images belonging to 5 classes.
In [5]: train_batches.class_indices
Out[5]: {'ApplyLipstick': 0,
         'Biking': 1,
         'Kayaking': 2,
         'ShavingBeard': 3,
         'TennisSwing': 4}
In [6]: from matplotlib import pyplot as plt
        def plot images(images arr):
            fig, axes = plt.subplots(1,10)
            axes = axes.flatten()
            for img, ax in zip(images arr, axes):
                ax.imshow(img)
                ax.axis('off')
            plt.tight layout()
            plt.show()
In [7]: imgs, labels = train batches[0]
        plot images(imgs)
        print(labels[:10])
```

```
[[0. 0. 0. 1. 0.]

[0. 1. 0. 0. 0.]

[0. 1. 0. 0. 0.]

[0. 0. 0. 1. 0.]

[0. 0. 0. 0. 1.]

[1. 0. 0. 0. 0.]

[0. 0. 1. 0. 0.]

[0. 0. 0. 1. 0.]

[0. 1. 0. 0. 0.]

[0. 1. 0. 0. 0.]
```

Initialize model

2022-08-25 22:35:44.535753: I tensorflow/core/platform/cpu_feature_guard. cc:193] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operations: AVX2 FMA

To enable them in other operations, rebuild TensorFlow with the appropria te compiler flags.

2022-08-25 22:35:44.537213: I tensorflow/stream executor/cuda/cuda gpu ex ecutor.cc:975] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-08-25 22:35:44.537955: I tensorflow/stream executor/cuda/cuda gpu ex ecutor.cc:975] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-08-25 22:35:44.538278: I tensorflow/stream executor/cuda/cuda gpu ex ecutor.cc:975] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-08-25 22:35:44.926780: I tensorflow/stream executor/cuda/cuda gpu ex ecutor.cc:975] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-08-25 22:35:44.926922: I tensorflow/stream executor/cuda/cuda gpu ex ecutor.cc:975] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-08-25 22:35:44.927028: I tensorflow/stream executor/cuda/cuda gpu ex ecutor.cc:975] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-08-25 22:35:44.927152: I tensorflow/core/common runtime/gpu/gpu devi ce.cc:1532] Created device /job:localhost/replica:0/task:0/device:GPU:0 w ith 3368 MB memory: -> device: 0, name: NVIDIA GeForce GTX 1050 Ti, pci bus id: 0000:01:00.0, compute capability: 6.1

```
(0, 'input 1', [(None, 224, 224, 3)])
    'conv1', (None, 112, 112, 32))
    'conv1 bn', (None, 112, 112, 32))
    'conv1_relu', (None, 112, 112, 32))
   'conv_dw_1', (None, 112, 112, 32))
   'conv dw 1 bn', (None, 112, 112, 32))
(5,
   'conv_dw_1_relu', (None, 112, 112, 32))
(6,
   'conv pw 1', (None, 112, 112, 64))
    'conv_pw_1_bn', (None, 112, 112, 64))
(9, 'conv_pw_1_relu', (None, 112, 112, 64))
(10, 'conv_pad_2', (None, 113, 113, 64))
(11, 'conv dw 2', (None, 56, 56, 64))
(12, 'conv_dw_2_bn', (None, 56, 56, 64))
(13, 'conv_dw_2_relu', (None, 56, 56, 64))
(14, 'conv_pw_2', (None, 56, 56, 128))
    'conv pw 2 bn', (None, 56, 56, 128))
(15,
(16, 'conv pw 2 relu', (None, 56, 56, 128))
    'conv dw 3', (None, 56, 56, 128))
(18, 'conv dw 3_bn', (None, 56, 56, 128))
     'conv dw 3 relu', (None, 56, 56, 128))
(19,
     'conv_pw_3', (None, 56, 56, 128))
(20,
     'conv_pw_3_bn', (None, 56, 56, 128))
(21,
(22, 'conv pw 3 relu', (None, 56, 56, 128))
(23, 'conv_pad_4', (None, 57, 57, 128))
(24, 'conv dw 4', (None, 28, 28, 128))
(25, 'conv_dw_4_bn', (None, 28, 28, 128))
(26, 'conv_dw_4_relu', (None, 28, 28, 128))
     'conv pw 4', (None, 28, 28, 256))
(27,
     'conv pw 4 bn', (None, 28, 28, 256))
     'conv_pw_4_relu', (None, 28, 28, 256))
(30,
     'conv_dw_5', (None, 28, 28, 256))
     'conv dw 5 bn', (None, 28, 28, 256))
(31,
     'conv_dw_5_relu', (None, 28, 28, 256))
     'conv pw 5', (None, 28, 28, 256))
(33,
(34, 'conv pw 5 bn', (None, 28, 28, 256))
(35, 'conv pw 5 relu', (None, 28, 28, 256))
(36, 'conv_pad_6', (None, 29, 29, 256))
(37, 'conv dw 6', (None, 14, 14, 256))
(38, 'conv_dw_6_bn', (None, 14, 14, 256))
    'conv dw 6 relu', (None, 14, 14, 256))
(39,
     'conv pw 6', (None, 14, 14, 512))
     'conv pw 6 bn', (None, 14, 14, 512))
(41,
     'conv pw 6 relu', (None, 14, 14, 512))
(42,
     'conv_dw_7', (None, 14, 14, 512))
(44,
     'conv_dw_7_bn', (None, 14, 14, 512))
(45,
     'conv dw 7 relu', (None, 14, 14, 512))
     'conv pw 7', (None, 14, 14, 512))
(47, 'conv_pw_7_bn', (None, 14, 14, 512))
(48, 'conv_pw_7_relu', (None, 14, 14, 512))
(49, 'conv dw 8', (None, 14, 14, 512))
    'conv dw 8 bn', (None, 14, 14, 512))
(50,
    'conv dw 8 relu', (None, 14, 14, 512))
(51,
     'conv pw 8', (None, 14, 14, 512))
     'conv pw 8 bn', (None, 14, 14, 512))
(53,
     'conv pw 8 relu', (None, 14, 14, 512))
(54,
     'conv_dw_9', (None, 14, 14, 512))
     'conv_dw_9_bn', (None, 14, 14, 512))
(57, 'conv dw 9 relu', (None, 14, 14, 512))
(58, 'conv pw 9', (None, 14, 14, 512))
(59, 'conv pw 9 bn', (None, 14, 14, 512))
(60, 'conv_pw_9_relu', (None, 14, 14, 512))
    'conv dw 10', (None, 14, 14, 512))
(62, 'conv dw 10 bn', (None, 14, 14, 512))
```

```
(63, 'conv dw 10 relu', (None, 14, 14, 512))
(64, 'conv_pw_10', (None, 14, 14, 512))
(65, 'conv pw 10 bn', (None, 14, 14, 512))
(66, 'conv_pw_10_relu', (None, 14, 14, 512))
(67, 'conv_dw_11', (None, 14, 14, 512))
(68, 'conv dw 11 bn', (None, 14, 14, 512))
(69, 'conv dw 11 relu', (None, 14, 14, 512))
(70, 'conv pw 11', (None, 14, 14, 512))
(71, 'conv_pw_11_bn', (None, 14, 14, 512))
(72, 'conv_pw_11_relu', (None, 14, 14, 512))
(73, 'conv_pad_12', (None, 15, 15, 512))
(74, 'conv dw 12', (None, 7, 7, 512))
(75, 'conv_dw_12_bn', (None, 7, 7, 512))
(76, 'conv dw_12_relu', (None, 7, 7, 512))
(77, 'conv_pw_12', (None, 7, 7, 1024))
(78, 'conv_pw_12_bn', (None, 7, 7, 1024))
(79, 'conv pw 12 relu', (None, 7, 7, 1024))
(80, 'conv dw 13', (None, 7, 7, 1024))
(81, 'conv dw 13 bn', (None, 7, 7, 1024))
(82, 'conv dw 13 relu', (None, 7, 7, 1024))
(83, 'conv_pw_13', (None, 7, 7, 1024))
(84,
    'conv_pw_13_bn', (None, 7, 7, 1024))
(85, 'conv_pw_13_relu', (None, 7, 7, 1024))
(86, 'global average pooling2d', (None, 1, 1, 1024))
(87, 'dropout', (None, 1, 1, 1024))
(88, 'conv_preds', (None, 1, 1, 1000))
(89, 'reshape_2', (None, 1000))
(90, 'predictions', (None, 1000))
mobilenet model = mobilenet.MobileNet(include top=False,
                   input_shape=(224,224,3),
                   pooling='avg',classes=5,
                   weights='imagenet')
for (i,layer) in enumerate(mobilenet model.layers):
    layer.trainable = False
    print((i, layer.name, layer.output shape))
```

```
(0, 'input 2', [(None, 224, 224, 3)])
    'conv1', (None, 112, 112, 32))
    'conv1 bn', (None, 112, 112, 32))
    'conv1_relu', (None, 112, 112, 32))
   'conv_dw_1', (None, 112, 112, 32))
   'conv dw 1 bn', (None, 112, 112, 32))
   'conv_dw_1_relu', (None, 112, 112, 32))
(6,
   'conv pw 1', (None, 112, 112, 64))
    'conv_pw_1_bn', (None, 112, 112, 64))
(9, 'conv_pw_1_relu', (None, 112, 112, 64))
(10, 'conv_pad_2', (None, 113, 113, 64))
(11, 'conv dw 2', (None, 56, 56, 64))
(12, 'conv_dw_2_bn', (None, 56, 56, 64))
(13, 'conv_dw_2_relu', (None, 56, 56, 64))
(14, 'conv_pw_2', (None, 56, 56, 128))
    'conv pw 2 bn', (None, 56, 56, 128))
(15,
(16, 'conv pw 2 relu', (None, 56, 56, 128))
(17, 'conv dw 3', (None, 56, 56, 128))
(18, 'conv dw 3_bn', (None, 56, 56, 128))
     'conv dw 3 relu', (None, 56, 56, 128))
(19,
     'conv_pw_3', (None, 56, 56, 128))
(20,
     'conv_pw_3_bn', (None, 56, 56, 128))
(21,
(22, 'conv pw 3 relu', (None, 56, 56, 128))
(23, 'conv_pad_4', (None, 57, 57, 128))
(24, 'conv dw 4', (None, 28, 28, 128))
(25, 'conv_dw_4_bn', (None, 28, 28, 128))
(26, 'conv_dw_4_relu', (None, 28, 28, 128))
     'conv pw 4', (None, 28, 28, 256))
(27,
     'conv pw 4 bn', (None, 28, 28, 256))
     'conv_pw_4_relu', (None, 28, 28, 256))
(30,
     'conv_dw_5', (None, 28, 28, 256))
     'conv dw 5 bn', (None, 28, 28, 256))
(31,
     'conv_dw_5_relu', (None, 28, 28, 256))
     'conv pw 5', (None, 28, 28, 256))
(33,
(34, 'conv pw 5 bn', (None, 28, 28, 256))
(35, 'conv pw 5 relu', (None, 28, 28, 256))
(36, 'conv_pad_6', (None, 29, 29, 256))
(37, 'conv dw 6', (None, 14, 14, 256))
(38, 'conv_dw_6_bn', (None, 14, 14, 256))
    'conv dw 6 relu', (None, 14, 14, 256))
(39,
     'conv pw 6', (None, 14, 14, 512))
     'conv pw 6 bn', (None, 14, 14, 512))
(41,
     'conv pw 6 relu', (None, 14, 14, 512))
(42,
     'conv_dw_7', (None, 14, 14, 512))
(44.
     'conv_dw_7_bn', (None, 14, 14, 512))
     'conv dw 7 relu', (None, 14, 14, 512))
(45,
     'conv pw 7', (None, 14, 14, 512))
(47, 'conv_pw_7_bn', (None, 14, 14, 512))
(48, 'conv_pw_7_relu', (None, 14, 14, 512))
(49, 'conv dw 8', (None, 14, 14, 512))
    'conv dw 8 bn', (None, 14, 14, 512))
(50,
    'conv dw 8 relu', (None, 14, 14, 512))
(51,
     'conv pw 8', (None, 14, 14, 512))
     'conv pw 8 bn', (None, 14, 14, 512))
(53,
     'conv pw 8 relu', (None, 14, 14, 512))
(54,
     'conv_dw_9', (None, 14, 14, 512))
     'conv_dw_9_bn', (None, 14, 14, 512))
(57, 'conv dw 9 relu', (None, 14, 14, 512))
(58, 'conv pw 9', (None, 14, 14, 512))
(59, 'conv pw 9 bn', (None, 14, 14, 512))
(60, 'conv_pw_9_relu', (None, 14, 14, 512))
    'conv dw 10', (None, 14, 14, 512))
(62, 'conv dw 10 bn', (None, 14, 14, 512))
```

```
(63, 'conv dw 10 relu', (None, 14, 14, 512))
         (64, 'conv_pw_10', (None, 14, 14, 512))
         (65, 'conv pw 10_bn', (None, 14, 14, 512))
         (66, 'conv_pw_10_relu', (None, 14, 14, 512))
         (67, 'conv_dw_11', (None, 14, 14, 512))
         (68, 'conv dw 11 bn', (None, 14, 14, 512))
         (69, 'conv dw 11 relu', (None, 14, 14, 512))
         (70, 'conv pw 11', (None, 14, 14, 512))
         (71, 'conv_pw_11_bn', (None, 14, 14, 512))
         (72, 'conv_pw_11_relu', (None, 14, 14, 512))
         (73, 'conv_pad_12', (None, 15, 15, 512))
         (74, 'conv dw 12', (None, 7, 7, 512))
         (75, 'conv_dw_12_bn', (None, 7, 7, 512))
         (76, 'conv dw_12_relu', (None, 7, 7, 512))
         (77, 'conv_pw_12', (None, 7, 7, 1024))
         (78, 'conv_pw_12_bn', (None, 7, 7, 1024))
         (79, 'conv pw 12 relu', (None, 7, 7, 1024))
         (80, 'conv dw 13', (None, 7, 7, 1024))
         (81, 'conv dw 13_bn', (None, 7, 7, 1024))
         (82, 'conv dw 13 relu', (None, 7, 7, 1024))
         (83, 'conv_pw_13', (None, 7, 7, 1024))
         (84,
              'conv_pw_13_bn', (None, 7, 7, 1024))
         (85, 'conv_pw_13_relu', (None, 7, 7, 1024))
         (86, 'global average pooling2d 1', (None, 1024))
In [10]:
         model = models.Sequential()
         dense layer 1 = Dense(32, activation='relu')
         dense layer 2 = Dense(32, activation='relu')
         dense layer 2 = Dense(32, activation='relu')
         prediction_layer = Dense(5, activation='softmax')
         model.add(mobilenet model)
         model.add(dense layer 1)
         model.add(dense layer 2)
         model.add(prediction layer)
         model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
mobilenet_1.00_224 (Functional)	(None, 1024)	3228864
dense (Dense)	(None, 32)	32800
dense_2 (Dense)	(None, 32)	1056
dense_3 (Dense)	(None, 5)	165
=======================================		========

Total names: 2 262 005

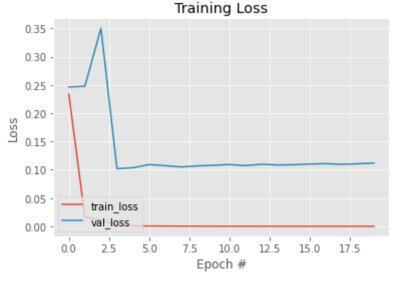
Total params: 3,262,885 Trainable params: 34,021

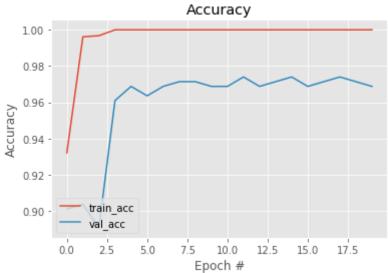
Non-trainable params: 3,228,864

```
accuracy: 0.9321 - val loss: 0.2462 - val accuracy: 0.9010
    Epoch 2/20
    accuracy: 0.9961 - val loss: 0.2476 - val accuracy: 0.9036
    Epoch 3/20
    accuracy: 0.9968 - val loss: 0.3501 - val accuracy: 0.8906
    Epoch 4/20
    accuracy: 1.0000 - val loss: 0.1018 - val accuracy: 0.9609
    Epoch 5/20
    4 - accuracy: 1.0000 - val loss: 0.1036 - val accuracy: 0.9688
    Epoch 6/20
    4 - accuracy: 1.0000 - val loss: 0.1092 - val accuracy: 0.9635
    Epoch 7/20
    4 - accuracy: 1.0000 - val loss: 0.1073 - val accuracy: 0.9688
    Epoch 8/20
    4 - accuracy: 1.0000 - val loss: 0.1049 - val accuracy: 0.9714
    Epoch 9/20
    4 - accuracy: 1.0000 - val loss: 0.1069 - val accuracy: 0.9714
    Epoch 10/20
    4 - accuracy: 1.0000 - val loss: 0.1079 - val accuracy: 0.9688
    Epoch 11/20
    4 - accuracy: 1.0000 - val loss: 0.1093 - val accuracy: 0.9688
    Epoch 12/20
    4 - accuracy: 1.0000 - val loss: 0.1073 - val accuracy: 0.9740
    Epoch 13/20
    4 - accuracy: 1.0000 - val loss: 0.1098 - val accuracy: 0.9688
    Epoch 14/20
    4 - accuracy: 1.0000 - val loss: 0.1085 - val accuracy: 0.9714
    Epoch 15/20
    4 - accuracy: 1.0000 - val loss: 0.1089 - val accuracy: 0.9740
    Epoch 16/20
    5 - accuracy: 1.0000 - val loss: 0.1101 - val accuracy: 0.9688
    Epoch 17/20
    5 - accuracy: 1.0000 - val loss: 0.1109 - val accuracy: 0.9714
    Epoch 18/20
    5 - accuracy: 1.0000 - val loss: 0.1096 - val accuracy: 0.9740
    Epoch 19/20
    5 - accuracy: 1.0000 - val loss: 0.1106 - val accuracy: 0.9714
    Epoch 20/20
    5 - accuracy: 1.0000 - val loss: 0.1119 - val accuracy: 0.9688
In [14]: model.save("./models/action-class-05-trained-mobilenet.h5")
```

Evaluate and Predict

```
model = models.load model("./models/action-class-05-trained-mobilenet.h5"
In [15]:
         model.summary()
         Model: "sequential"
         Layer (type)
                                    Output Shape
                                                             Param #
          mobilenet 1.00 224 (Functio (None, 1024)
                                                             3228864
          dense (Dense)
                                    (None, 32)
                                                             32800
         dense 2 (Dense)
                                    (None, 32)
                                                             1056
          dense 3 (Dense)
                                    (None, 5)
                                                             165
         Total params: 3,262,885
        Trainable params: 34,021
         Non-trainable params: 3,228,864
In [16]:
        model.evaluate(test_batches)
         curacy: 0.9688
        [0.11185693740844727, 0.96875]
Out[16]:
In [17]:
        plt.style.use("ggplot")
         plt.figure()
         plt.plot(np.arange(0, 20), fit.history["loss"], label="train_loss")
         plt.plot(np.arange(0, 20), fit.history["val loss"], label="val loss")
         plt.title("Training Loss")
         plt.xlabel("Epoch #")
         plt.ylabel("Loss")
         plt.legend(loc="lower left")
         plt.show()
         plt.plot(np.arange(0, 20), fit.history["accuracy"], label="train_acc")
         plt.plot(np.arange(0, 20), fit.history["val_accuracy"], label="val acc")
         plt.title("Accuracy")
         plt.xlabel("Epoch #")
         plt.ylabel("Accuracy")
         plt.legend(loc="lower left")
         plt.show()
```





```
In [18]: print("Avg Val Acc: " + str(sum(fit.history["val_accuracy"])/20*100))
print("Avg Val Loss: " + str(sum(fit.history["val_loss"])/20*100))
```

Avg Val Acc: 95.89843779802322 Avg Val Loss: 13.41185424476862