fine-tune-2 class

January 23, 2020

[9]: from tensorflow.keras import applications

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
    from tensorflow.keras import optimizers
    from tensorflow.keras.models import Sequential
    from tensorflow.keras.layers import Dropout, Flatten, Dense
    from tensorflow.keras.layers import Input
    from tensorflow.keras.models import load_model
    from tensorflow.keras.models import Model
    from tensorflow.keras.utils import plot_model
    from sklearn.metrics import classification_report
    import matplotlib.pyplot as plt
    import numpy as np
    import os
    import tensorflow as tf
[4]: tf.config.threading.set_inter_op_parallelism_threads(6)
    tf.config.threading.set_intra_op_parallelism_threads(2)
     # path to the model weights files.
    weights_path = '/home/user/models/pretrained/vgg16_weights.h5'
    top_model_weights_path = '/home/user/models/top_tuned/
     # dimensions of our images.
    img_width, img_height = 224, 224
    train_data_dir = '/home/user/
                                      /convnets/transfer-learning-keras/dataset/
     →training'
    validation_data_dir = '/home/user/
                                           /convnets/transfer-learning-keras/

→dataset/validation'

    evaluation_data_dir = '/home/user/
                                           /convnets/transfer-learning-keras/

→dataset/evaluation'

    nb train samples = 3000
    nb_validation_samples = 1000
    nb_evaluation_samples = 1000
    epochs = 10
    batch_size = 20
```

VGG16 model loaded.

```
[5]: # build a classifier model to put on top of the convolutional model
     top_model = base_model.output
     top model = Flatten(name='flatten')(top model)
     top_model = Dense(512, activation='relu')(top_model)
     top_model = Dropout(0.5)(top_model)
     top_model = Dense(1, activation='sigmoid')(top_model)
     # note that it is necessary to start with a fully-trained
     # classifier, including the top classifier,
     # in order to successfully do fine-tuning
     # add the model on top of the convolutional base
     model = Model(inputs=base_model.input, outputs=top_model)
     # set the first 15 layers (up to the last conv block)
     # to non-trainable (weights will not be updated)
     for layer in base_model.layers:
         layer.trainable = False
     for layer in model.layers:
         print("{}: {}".format(layer, layer.trainable))
    <tensorflow.python.keras.engine.input_layer.InputLayer object at</pre>
    0x7f45bc2b58d0>: False
    <tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc323d50>:
    <tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc2b5f10>:
    <tensorflow.python.keras.layers.pooling.MaxPooling2D object at 0x7f45bc2bf2d0>:
    False
    <tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc2c4990>:
    False
    <tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc2d25d0>:
    <tensorflow.python.keras.layers.pooling.MaxPooling2D object at 0x7f45bc2e20d0>:
    <tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc2e8e90>:
    <tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc276450>:
    <tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc280750>:
    False
```

```
<tensorflow.python.keras.layers.pooling.MaxPooling2D object at 0x7f45bc292650>:
False
<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc29f2d0>:
False
<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc2a5bd0>:
False
<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc1f6a90>:
False
<tensorflow.python.keras.layers.pooling.MaxPooling2D object at 0x7f45bc209590>:
False
<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc20f390>:
False
<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc21c990>:
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<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc1ae550>:
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<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc1ae550>:
False
<tensorflow.python.keras.layers.pooling.MaxPooling2D object at 0x7f45bc1ae550>:
False
<tensorflow.python.keras.layers.core.Flatten object at 0x7f45bc194510>: True
<tensorflow.python.keras.layers.core.Dense object at 0x7f45bc18dcd0>: True
<tensorflow.python.keras.layers.core.Dense object at 0x7f45bc18db10>: True
<tensorflow.python.keras.layers.core.Dense object at 0x7f45bc18db10>: True
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```

```
[6]: # compile the model with a SGD/momentum optimizer
     # and a very slow learning rate.
     model.compile(loss='binary_crossentropy',
                   optimizer=optimizers.SGD(lr=1e-4, momentum=0.9),
                   metrics=['accuracy'])
     # prepare data augmentation configuration
     train_datagen = ImageDataGenerator(
             rotation_range=40,
             width_shift_range=0.2,
             height_shift_range=0.2,
             shear_range=0.2,
             zoom_range=0.2,
             horizontal_flip=True,
             fill_mode='nearest'
     )
     test_datagen = ImageDataGenerator()
     mean = np.array([123.68, 116.779, 103.939], dtype="float32")
     train_datagen.mean = mean
     test_datagen.mean = mean
     train_generator = train_datagen.flow_from_directory(
         train_data_dir,
```

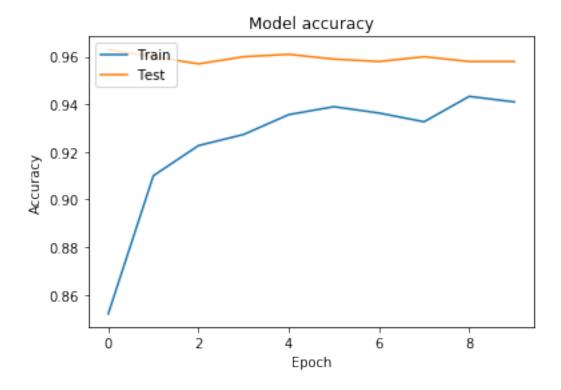
```
color_mode="rgb",
   target_size=(img_height, img_width),
   batch_size=batch_size,
   shuffle=True,
   class_mode='binary')
validation_generator = test_datagen.flow_from_directory(
   validation_data_dir,
   color mode="rgb",
   target_size=(img_height, img_width),
   batch size=batch size,
   shuffle=False,
   class_mode='binary')
test_generator = test_datagen.flow_from_directory(
   evaluation_data_dir,
   color_mode="rgb",
   target_size=(img_height, img_width),
   batch_size=batch_size,
   shuffle=False,
   class_mode='binary')
print("[INFO] training head...")
H = model.fit generator(
   train_generator,
   steps_per_epoch=nb_train_samples // batch_size,
   epochs=epochs,
   validation_data=validation_generator,
   validation_steps=nb_validation_samples // batch_size)
Found 3000 images belonging to 2 classes.
Found 1000 images belonging to 2 classes.
Found 1000 images belonging to 2 classes.
[INFO] training head...
Epoch 1/10
0.8513Epoch 1/10
0.8520 - val_loss: 0.4303 - val_acc: 0.9630
Epoch 2/10
0.9101Epoch 1/10
0.9100 - val_loss: 0.3328 - val_acc: 0.9600
Epoch 3/10
0.9228Epoch 1/10
```

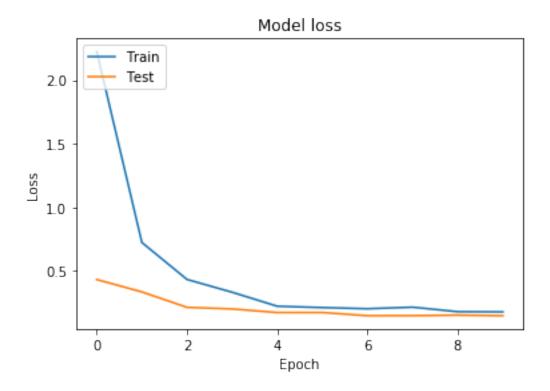
```
0.9227 - val_loss: 0.2112 - val_acc: 0.9570
  Epoch 4/10
  0.9272Epoch 1/10
  0.9273 - val_loss: 0.1986 - val_acc: 0.9600
  Epoch 5/10
  0.9359Epoch 1/10
  0.9357 - val_loss: 0.1703 - val_acc: 0.9610
  Epoch 6/10
  0.9393Epoch 1/10
  0.9390 - val_loss: 0.1706 - val_acc: 0.9590
  Epoch 7/10
  0.9366Epoch 1/10
  0.9363 - val_loss: 0.1454 - val_acc: 0.9580
  Epoch 8/10
  0.9329Epoch 1/10
  0.9327 - val_loss: 0.1459 - val_acc: 0.9600
  Epoch 9/10
  0.9433Epoch 1/10
  0.9433 - val_loss: 0.1499 - val_acc: 0.9580
  Epoch 10/10
  0.9409Epoch 1/10
  0.9410 - val_loss: 0.1452 - val_acc: 0.9580
[7]: model.save_weights('/home/user/models/fine_tuned/fine_tuned_2class.h5')
[11]: plot_model(model, to_file='model.png')
    # Plot training & validation accuracy values
  plt.plot(H.history['acc'])
  plt.plot(H.history['val_acc'])
  plt.title('Model accuracy')
  plt.ylabel('Accuracy')
  plt.xlabel('Epoch')
```

```
plt.legend(['Train', 'Test'], loc='upper left')
plt.show()

# Plot training & validation loss values
plt.plot(H.history['loss'])
plt.plot(H.history['val_loss'])
plt.title('Model loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['Train', 'Test'], loc='upper left')
plt.show()
```

Failed to import pydot. You must install pydot and graphviz for `pydotprint` to work.





```
[13]: for layer in base_model.layers[15:]:
    layer.trainable = True
for layer in model.layers:
    print("{}: {}".format(layer, layer.trainable))
```

 $\verb| <tensorflow.python.keras.engine.input_layer.InputLayer object at | \\$

0x7f45bc2b58d0>: False

<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc323d50>:
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<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc2b5f10>:
False

<tensorflow.python.keras.layers.pooling.MaxPooling2D object at 0x7f45bc2bf2d0>:
Folso

<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc2c4990>:
False

<tensorflow.python.keras.layers.pooling.MaxPooling2D object at 0x7f45bc2e20d0>:
False

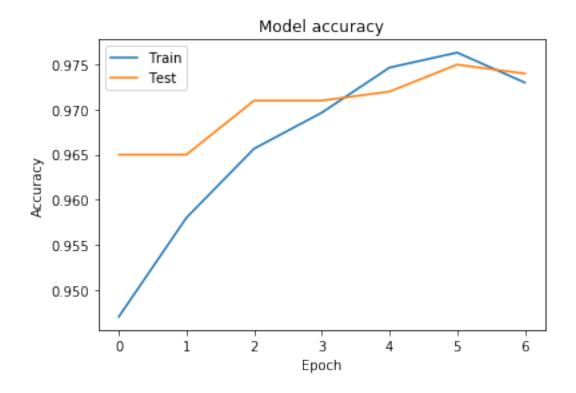
 $\verb|\convolutional.Conv2D| object at 0x7f45bc2e8e90>: False \\$

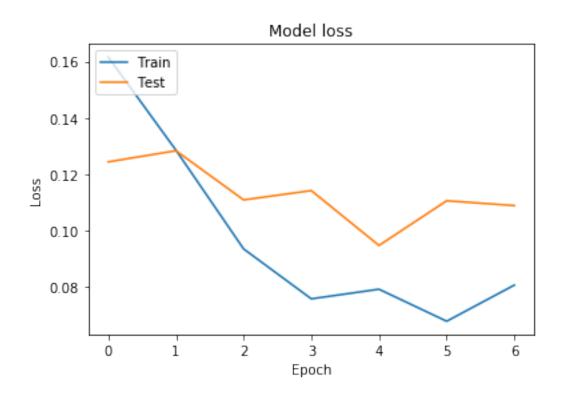
<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc276450>:
False

<tensorflow.python.keras.layers.convolutional.Conv2D object at 0x7f45bc280750>:

```
False
    <tensorflow.python.keras.layers.pooling.MaxPooling2D object at 0x7f45bc292650>:
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    <tensorflow.python.keras.layers.pooling.MaxPooling2D object at 0x7f45bc209590>:
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    <tensorflow.python.keras.layers.pooling.MaxPooling2D object at 0x7f45bc1c0090>:
    True
    <tensorflow.python.keras.layers.core.Flatten object at 0x7f45bc194510>: True
    <tensorflow.python.keras.layers.core.Dense object at 0x7f45bc18dcd0>: True
    <tensorflow.python.keras.layers.core.Dropout object at 0x7f45bc18db10>: True
    <tensorflow.python.keras.layers.core.Dense object at 0x7f45bc18dd10>: True
[16]: train_generator.reset()
     validation_generator.reset()
     print("[INFO] re-compiling model...")
     opt = optimizers.SGD(lr=1e-4, momentum=0.9)
     model.compile(loss="binary_crossentropy", optimizer=opt,
        metrics=["accuracy"])
     print("[INFO] training with last conv block...")
     H = model.fit_generator(
        train_generator,
        steps_per_epoch=nb_train_samples // batch_size,
        epochs=7,
        validation_data=validation_generator,
        validation_steps=nb_validation_samples // batch_size)
    [INFO] re-compiling model...
    [INFO] training with last conv block...
    Epoch 1/7
    0.9470Epoch 1/7
    0.9470 - val_loss: 0.1243 - val_acc: 0.9650
    Epoch 2/7
    0.9591Epoch 1/7
```

```
0.9580 - val_loss: 0.1283 - val_acc: 0.9650
   Epoch 3/7
   0.9654Epoch 1/7
   0.9657 - val loss: 0.1108 - val acc: 0.9710
   Epoch 4/7
   0.9701Epoch 1/7
   0.9697 - val_loss: 0.1141 - val_acc: 0.9710
   Epoch 5/7
   0.9748Epoch 1/7
   0.9747 - val_loss: 0.0946 - val_acc: 0.9720
   Epoch 6/7
   0.9762Epoch 1/7
   0.9763 - val_loss: 0.1105 - val_acc: 0.9750
   Epoch 7/7
   0.9728Epoch 1/7
   0.9730 - val_loss: 0.1088 - val_acc: 0.9740
[17]: model.save_weights('/home/user/models/fine_tuned/fine_tuned_2class_full.h5')
     # Plot training & validation accuracy values
   plt.plot(H.history['acc'])
   plt.plot(H.history['val_acc'])
   plt.title('Model accuracy')
   plt.ylabel('Accuracy')
   plt.xlabel('Epoch')
   plt.legend(['Train', 'Test'], loc='upper left')
   plt.show()
     # Plot training & validation loss values
   plt.plot(H.history['loss'])
   plt.plot(H.history['val_loss'])
   plt.title('Model loss')
   plt.ylabel('Loss')
   plt.xlabel('Epoch')
   plt.legend(['Train', 'Test'], loc='upper left')
   plt.show()
```





[]:[