```
ATTENTION:
             Please
                    do not alter any of the provided code in the exercise. Only add
             Please
                    do not add or remove any cells in the exercise. The grader will
  ATTENTION:
            Please use the provided epoch values when training.
  ATTENTION:
  Import all the necessary files!
import os
import tensorflow as tf
from tensorflow.keras import layers
from tensorflow.keras import Model
from os import getcwd
path_inception = f"{getcwd()}/../tmp2/inception_v3_weights_tf_dim_ordering_tf_kernels_notop.h5"
# Import the inception model
from tensorflow.keras.applications.inception v3 import InceptionV3
# Create an instance of the inception model from the local pre-trained weights
local_weights_file = path_inception
pre trained model = # Your Code Here
pre_trained_model.load_weights(local_weights_file)
# Make all the layers in the pre-trained model non-trainable
for layer in pre_trained_model.layers:
   # Your Code Here
# Print the model summary
pre trained model. summary()
# Expected Output is extremely large, but should end with:
#batch normalization v1 281 (Bat (None, 3,
                                                        576
                                                                           conv2d 281[0][0
                                         3,
                                             192)
#activation 273 (Activation)
                                   (None,
                                          3,
                                             3,
                                                 320)
                                                           0
                                                                                batch_nor
#mixed9 1 (Concatenate)
                                                                0
                                                  3,
                                                     768)
                                       (None,
                                              3,
                                                                                    acti
#concatenate 5 (Concatenate)
                                   (None,
                                          3.
                                             3.
                                                 768)
                                                           0
                                                                                activatio
#activation 281 (Activation)
                                   (None,
                                          3.
                                             3.
                                                 192)
                                                           0
                                                                                batch nor
#mixed10 (Concatenate)
                                        (None,
                                              3,
                                                  3,
                                                      2048)
                                                                0
                                                                                    acti
#
#-----
```

#Total params: 21,802,784
#Trainable params: 0

#Non-trainable params: 21,802,784

```
last_layer = pre_trained_model.get_layer(# Your Code Here)
print('last layer output shape: ', last_layer.output_shape)
last output = # Your Code Here
# Expected Output:
  ('last layer output shape: ', (None, 7, 7,
                                                 768))
  Define a Callback class that stops training once accuracy reaches 97.0%
class myCallback(tf.keras.callbacks.Callback):
   def on epoch end(self, epoch, logs={}):
       if (logs. get ('acc')>0.97):
           print("\nReached 97.0% accuracy so cancelling training!")
           self.model.stop training = True
from tensorflow.keras.optimizers import RMSprop
  Flatten the output layer to 1 dimension
  = layers. Flatten() (last output)
  Add a fully connected layer with 1,024 hidden units and ReLU activation
  = layers. Dense (# Your Code Here) (x)
  Add a dropout rate of 0.2
  = layers. Dropout (# Your Code Here) (x)
  Add a final sigmoid layer for classification
  = layers. Dense (# Your Code Here)(x)
model = Model( # Your Code Here, x)
model.compile(optimizer = RMSprop(lr=0.0001),
                         loss = # Your Code Here,
                         metrics = # Your Code Here)
model.summary()
  Expected output will be large. Last few lines should be:
  mixed7 (Concatenate)
                                                                     0
                                            (None, 7, 7,
                                                          768)
#
#
#
#
#
  flatten 4 (Flatten)
                                             (None.
                                                    37632)
                                                                        0
#
#
  dense_8 (Dense)
                                                 (None,
                                                      1024)
                                                                             38536192
#
  dropout 4 (Dropout)
                                             (None,
                                                    1024)
                                                                         0
#
#
  dense 9 (Dense)
                                                      1)
                                                                                1025
                                                 (None,
  _____
#
  Total params: 47,512,481
  Trainable params: 38,537,217
  Non-trainable params: 8,975,264
```

```
# Get the Horse or Human dataset
path horse or human = f"{getcwd()}/../tmp2/horse-or-human.zip"
# Get the Horse or Human Validation dataset
path_validation_horse_or_human = f"{getcwd()}/../tmp2/validation-horse-or-human.zip"
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import os
import zipfile
import shutil
shutil.rmtree('/tmp')
local_zip = path_horse_or_human
zip_ref = zipfile.ZipFile(local_zip,
zip ref.extractall('/tmp/training')
zip ref. close()
local zip = path validation horse or human
zip ref = zipfile. ZipFile(local zip,
zip_ref. extractall('/tmp/validation')
zip ref.close()
# Define our example directories and files
train_dir = '/tmp/training'
validation dir = '/tmp/validation'
train_horses_dir = # Your Code Here
train humans dir = # Your Code Here
validation horses dir = # Your Code Here
validation_humans_dir = # Your Code Here
train horses fnames = # Your Code Here
train humans fnames = # Your Code Here
validation_horses_fnames = # Your Code Here
validation humans fnames = # Your Code Here
print(# Your
             Code Here)
print(#
        Your
              Code Here)
print(#
              Code Here)
        Your
print(# Your Code Here)
  Expected Output:
  500
#
  527
#
  128
#
  128
  Add our data-augmentation parameters to ImageDataGenerator
train datagen = ImageDataGenerator(# Your Code Here)
# Note that the validation data should not be augmented!
```

test datagen = ImageDataGenerator(# Your Code Here)

```
# Flow training images in batches of 20 using train_datagen generator
train generator = train datagen. flow from directory (# Your Code Here)
# Flow validation images in batches of 20 using test datagen generator
validation_generator = test_datagen.flow_from_directory( # Your Code Here)
 Expected Output:
# Found 1027 images belonging to 2 classes.
# Found 256 images belonging to 2 classes.
  Run this and see how many epochs it should take before the callback
  fires, and stops training at 97% accuracy
callbacks = # Your Code Here
history = model.fit generator(# Your Code Here (set epochs = 3))
%matplotlib inline
import matplotlib.pyplot as plt
acc = history.history['acc']
val acc = history.history['val acc']
loss = history.history['loss']
val loss = history.history['val loss']
epochs = range(len(acc))
plt.plot(epochs, acc, 'r', label='Training accuracy')
plt.plot(epochs, val_acc, 'b', label='Validation accuracy')
plt.title('Training and validation accuracy')
plt.legend(loc=0)
plt.figure()
plt.show()
```

Submission Instructions

```
# Now click the 'Submit Assignment' button above.
```

When you're done or would like to take a break, please run the two cells below to save your work and close the Notebook. This will free up resources for your fellow learners.

```
%%javascript
<!-- Save the notebook -->
IPython.notebook.save_checkpoint();

%%javascript
IPython.notebook.session.delete();
window.onbeforeunload = null
setTimeout(function() { window.close(); }, 1000);
```