```
 1 import tensorflow as tf
 2 from tensorflow.keras.callbacks import
   LearningRateScheduler
 3 from tensorflow.keras.layers import Dense,  Conv2D,
   MaxPooling2D, Dropout, Flatten,
   GlobalAveragePooling2D
 4 from tensorflow.keras.optimizers import Adam, SGD,
   RMSprop
 5 from tensorflow.keras.preprocessing.image import
   ImageDataGenerator
 6 from tensorflow.keras.callbacks import Callback
 7 import numpy as np
 8 import matplotlib.pyplot as plt
 9 import os
10 from os.path import join
11 import multiprocessing
12
13 ImageResolution = (640, 360)
14 ImageResolutionGrayScale = (640, 360, 1)
15
16 def trainCNN( ):
17
18     tf.keras.backend.clear_session()
19
20     base_dir = 'G:\GIEyA\TFG\meteor_classification\
   labeledData\evenData'
21     results_dir_weights = 'G:\GIEyA\TFG\
   meteor_classification\\results\weights'
22
23     train_dir = join(base_dir, 'train')
24     validation_dir = join(base_dir, 'valid')
25
26     train_meteors_dir = join(train_dir, 'meteors')
27     train_non_meteors_dir = join(train_dir, '
   non_meteors')
28     validation_meteors_dir = join(validation_dir, '
   meteors')
29     validation_non_meteors_dir = join(validation_dir
   , 'non_meteors')
30
31     print('total training meteors images: ', len(os.
   listdir(train_meteors_dir)))
32     print('total training non-meteors images: ', len(
   os.listdir(train_non_meteors_dir)))
```

```
33      print('total validation meteors images: ', len(os
   .listdir(validation_meteors_dir)))
34      print('total validation non-meteors images: ',
   len(os.listdir(validation_non_meteors_dir)))
35
36
37      #Rescale all images by 1./255
38
39      train_datagen = ImageDataGenerator(rescale=1.0/
   255#,
40                                         #
   rotation_range=10, # Range from 0 to 180 degrees to
   randomly rotate images
41                                         #
   width_shift_range=0.05,
42                                         #
   height_shift_range=0.05,
43                                         #shear_range=5
   , # Shear the image by 5 degrees
44                                         #zoom_range=0.
   1,
45                                         #
   horizontal_flip=True,
46                                         #vertical_flip
   =True,
47                                         #fill_mode='
   nearest'
48                                         )
49
50      test_datagen = ImageDataGenerator(rescale=1.0/255
   .)
51
52      train_generator = train_datagen.
   flow_from_directory(train_dir,
53
      batch_size=16, #16
54
      class_mode='binary',
55
      color_mode='grayscale',
56
      target_size=ImageResolution) # 640x360 = 480x480
   . (640, 360)
57      validation_generator = test_datagen.
```

```
57 flow_from_directory(validation_dir,
58 
           batch_size=16, #16
59 
           class_mode='binary',
60 
           color_mode='grayscale',
61 
           target_size=ImageResolution)
62 
63 
64     model = tf.keras.models.Sequential([
65         Conv2D(64, (7, 7), activation='relu',
   input_shape=ImageResolutionGrayScale, strides=2),
66         MaxPooling2D(pool_size=(3,3), strides=2),
67         Dropout(0.25),
68 
69         Conv2D(32, (3, 3), activation='relu',
   kernel_initializer='he_uniform'),
70         Conv2D(32, (3, 3), activation='relu',
   kernel_initializer='he_uniform'),
71         MaxPooling2D(pool_size=(3, 3)),
72         Dropout(0.25),
73 
74         Conv2D(16, (3, 3), activation='relu',
   kernel_initializer='he_uniform'),
75         Conv2D(16, (3, 3), activation='relu',
   kernel_initializer='he_uniform'),
76         MaxPooling2D(pool_size=(3, 3)),
77         Dropout(0.25),
78 
79         Conv2D(8, (3, 3), activation='relu',
   kernel_initializer='he_uniform'),
80         Conv2D(8, (3, 3), activation='relu',
   kernel_initializer='he_uniform'),
81         MaxPooling2D(pool_size=(3, 3)),
82 
83         Flatten(),
84         Dense(24, activation='relu',
   kernel_initializer='he_uniform'),
85         Dropout(0.25),
86         Dense(8, activation='relu',
   kernel_initializer='he_uniform'),
87         Dropout(0.25),
```

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 88          Dense(1, activation='sigmoid',
    kernel_initializer='he_uniform')
 89      ])
 90
 91      print(model.summary())
 92      optimizer = Adam(learning_rate=5e-4) #3e-3 # Try
     with more and less learning rate # 5e-3
 93      model.compile(optimizer=optimizer,
 94                    loss='binary_crossentropy',
 95                    metrics=['accuracy'])
 96      #model.load_weights(join(results_dir_weights, '
    model_acc_0.926.h5'))
 97
 98      class SaveModelCallback(Callback):
 99          def __init__(self, threshold):
100              super(SaveModelCallback, self).__init__
    ()
101              self.threshold = threshold
102
103          def on_epoch_end(self, epoch, logs=None):
104              if(logs.get('accuracy') >= self.
    threshold):
105                  model.save_weights(join(
    results_dir_weights, '10_layers_model_2_acc_' +  str
    (logs.get('accuracy'))[0:6] + '_val_acc' + str(logs.
    get('val_accuracy'))[0:6] + '.h5'), save_format='h5'
    )
106
107      callback90 = SaveModelCallback(0.900)
108
109      #39.480 -> Training 39480 = 2 x 2 x 2 × 3 × 5 ×
    7 × 47
110      #9.872 -> Validation = 2 x 2 x 2 x 2 × 617
111      history = model.fit(train_generator,
112                          validation_data=
    validation_generator,
113                          steps_per_epoch=2467, #2467
     #4934
114                          epochs=150, #Later train
    with more epochs if neccessary
115                          validation_steps=617, #617 #
    1234
116                          verbose=1,
117                          callbacks=[callback90])
```

```
118
119     acc      = history.history['accuracy']
120     val_acc  = history.history['val_accuracy']
121     loss     = history.history['loss']
122     val_loss = history.history['val_loss']
123     epochs = range(len(acc)) #Get number of epochs
124
125     plt.plot(epochs, acc)
126     plt.plot(epochs, val_acc)
127     plt.title('Meteor detection training and
    validation accuracy')
128
129     plt.figure()
130     plt.plot(epochs, loss)
131     plt.plot(epochs, val_loss)
132     plt.title('Meteor detection training and
    validation loss')
133
134     plt.show()
135
136 if __name__ == '__main__':
137     p = multiprocessing.Process(target=trainCNN)
138     p.start()
139     p.join()
140
141
```