```
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\model_18'
22
23
       train_dir = join(base_dir, 'train')
       validation_dir = join(base_dir, 'valid')
24
25
       train_meteors_dir = join(train_dir, 'meteors')
26
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)))
```

```
print('total validation meteors images: ', len(os
33
   .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
       train_datagen = ImageDataGenerator(rescale=1.0/
39
   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
       model = tf.keras.models.Sequential([
64
           Conv2D(16, (11, 11), activation='relu',
65
   input_shape=ImageResolutionGrayScale, strides=2),
           MaxPooling2D(pool_size=(3, 3)),
66
           Dropout(0.25),
67
68
69
           Conv2D(12, (7, 7), activation='relu',
   kernel_initializer='he_uniform'),
           #Conv2D(12, (3, 3), activation='relu',
70
   kernel_initializer='he_uniform'),
71
           MaxPooling2D(pool_size=(2, 2)),
72
           Dropout(0.25),
73
           Conv2D(12, (5, 5), activation='relu',
74
   kernel_initializer='he_uniform'),
75
           #Conv2D(16, (3, 3), activation='relu',
   kernel_initializer='he_uniform'),
           MaxPooling2D(pool_size=(2, 2)),
76
77
           Dropout(0.25),
78
           Conv2D(12, (3, 3), activation='relu',
79
   kernel_initializer='he_uniform'),
           #Conv2D(8, (3, 3), activation='relu',
80
   kernel_initializer='he_uniform'),
           MaxPooling2D(pool_size=(2, 2)),
81
82
83
           Flatten(),
84
           Dense(480, activation='relu',
   kernel_initializer='he_uniform'),
85
           Dropout(0.30),
           Dense(16, activation='relu',
86
   kernel_initializer='he_uniform'),
           Dropout(0.20),
87
```

```
Dense(1, activation='sigmoid',
 88
    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',
                        metrics=['accuracy'])
 95
        #model.load_weights(join(results_dir_weights, '
 96
    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
             def on_epoch_end(self, epoch, logs=None):
103
                 if(logs.get('accuracy') >= self.
104
    threshold):
105
                      model.save_weights(join(
    results_dir_weights, '10_layers_model_2_acc_' +
    (logs.get('accuracy'))[0:6] + '_val_acc' + str(logs.
    get('val_accuracy'))[0:6] + '.h5'), save_format='h5'
106
        callback92 = SaveModelCallback(0.920)
107
108
109
        \#39.480 \rightarrow \text{Training } 39480 = 2 \times 2 \times 2 \times 3 \times 5 \times 6
    7 \times 47
110
        \#9.872 \rightarrow Validation = 2 \times 2 \times 2 \times 2 \times 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=[callback92])
```

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