```
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 from sklearn.metrics import confusion_matrix
13
14 def trainCNN():
15
16
       tf.keras.backend.clear_session()
17
18
       ImageResolution = (640, 360)
19
       ImageResolutionGrayScale = (640, 360, 1)
20
       modelNumber = 'Model21'
21
22
       base_dir = 'C:\work_dir\meteorData\extraData'
23
       results_dir_weights = 'G:\GIEyA\TFG\
   meteor_classification\\results\weights\\' +
   modelNumber
24
25
       train_dir = join(base_dir, 'train')
       validation_dir = join(base_dir, 'validation')
26
27
       test_dir = join(base_dir, 'test')
28
       #Rescale all images by 1./255
29
30
31
       train_datagen = ImageDataGenerator(rescale=1.0/
   255#,
32
   rotation_range=10, # Range from 0 to 180 degrees to
   randomly rotate images
33
                                           #
   width_shift_range=0.05,
```

```
34
                                            #
   height_shift_range=0.05,
35
                                            #shear_range=5
   , # Shear the image by 5 degrees
36
                                            #zoom_range=0.
   1,
37
                                            #
   horizontal_flip=True,
38
                                            #vertical_flip
   =True,
39
                                            #fill_mode='
   nearest'
40
                                            )
41
42
       validation_datagen = ImageDataGenerator(rescale=1
   .0/255.)
43
       test_datagen = ImageDataGenerator(rescale=1.0/255
   .0)
44
45
       train_generator = train_datagen.
   flow_from_directory(train_dir,
46
      batch_size=16, #16
47
      class_mode='binary',
48
      color_mode='grayscale',
49
      target_size=ImageResolution) # 640x360 = 480x480
   . (640, 360)
50
51
       validation_generator = validation_datagen.
   flow_from_directory(validation_dir,
52
                batch_size=16, #16
53
                class_mode='binary',
54
                color_mode='grayscale',
55
                target_size=ImageResolution)
56
57
       test_generator = test_datagen.flow_from_directory
   (test_dir,
```

```
58
     batch_size=1,
59
     class_mode='binary',
60
     color_mode='grayscale',
61
     target_size=ImageResolution,
62
     shuffle=False)
63
       model = tf.keras.models.Sequential([
64
           Conv2D(8, (11, 11), activation='relu',
65
   input_shape=ImageResolutionGrayScale, strides=1),
           MaxPooling2D(pool_size=(3, 3)),
66
           Dropout(0.25),
67
68
69
           Conv2D(16, (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)),
           Dropout(0.25),
72
73
           Conv2D(16, (3, 3), 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
79
           Conv2D(24, (3, 3), activation='relu',
   kernel_initializer='he_uniform'),
80
           MaxPooling2D(pool_size=(2, 2)),
81
           Conv2D(24, (3, 3), activation='relu',
82
   kernel_initializer='he_uniform'),
           MaxPooling2D(pool_size=(2, 2)),
83
84
85
           Flatten(),
           Dense(1320, activation='relu',
86
   kernel_initializer='he_uniform'),
           Dropout(0.30),
87
           Dense(512, activation='relu',
88
```

```
88 kernel_initializer='he_uniform'),
 89
            Dropout(0.20),
            Dense(1, activation='sigmoid',
 90
   kernel_initializer='he_uniform')
 91
        1)
 92
 93
        print(model.summary())
 94
        optimizer = Adam(learning_rate=5e-4) #3e-3 # Try
     with more and less learning rate # 5e-3
        model.compile(optimizer=optimizer,
 95
 96
                      loss='binary_crossentropy',
                      metrics=['accuracy'])
 97
        #model.load_weights(join(results_dir_weights, '
 98
    model_acc_0.926.h5'))
 99
100
        class SaveModelCallback(Callback):
            def __init__(self, thresholdTrain,
101
    thresholdValid):
102
                super(SaveModelCallback, self).__init__
    ()
103
                self.thresholdTrain = thresholdTrain
104
                self.thresholdValid = thresholdValid
105
            def on_epoch_end(self, epoch, logs=None):
106
                if((logs.get('accuracy') >= self.
107
    thresholdTrain) and (logs.get('val_accuracy') >=
    self.thresholdValid)):
                    model.save_weights(join(
108
    results_dir_weights, modelNumber + '_acc_' + str(
    logs.get('accuracy'))[0:6] + '_val_acc' + str(logs.
    get('val_accuracy'))[0:6] + '.h5'), save_format='h5'
109
110
        callback90 = SaveModelCallback(0.900, 0.900)
111
112
        # Training -> 66947
        # Validation -> 13388
113
114
        # Test -> 8928
115
116
        history = model.fit(train_generator,
117
                             validation_data=
    validation_generator,
118
                             steps_per_epoch=4184, #4184
119
                             epochs=50, #Later train with
```

```
119
    more epochs if neccessary
120
                        validation_steps=836, #836
121
                        verbose=1,
122
                        callbacks=[callback90])
123
124
   125
   126
       prob_predicted = model.predict(test_generator,
127
   steps=len(test_generator.filenames))
128
       test labels = []
129
      for i in range(0, len(test_generator.filenames
130
   )):
131
          test_labels.extend(np.array(test_generator[i
   ][1]))
132
      # Get the confusion matrix:
133
134
      truePositives = 0
135
      trueNegatives = 0
136
      falsePositives = 0
137
      falseNegatives = 0
138
139
      for i in range(len(prob_predicted)):
140
          if(prob_predicted[i] >= 0.5 and test_labels[
   il == 1.0):
141
             truePositives += 1
142
          elif(prob_predicted[i] >= 0.5 and
   test_labels[i] == 0.0):
143
             falsePositives += 1
144
          elif(test_labels[i] == 0.0):
145
             trueNegatives += 1
          elif(test_labels[i] == 1.0):
146
147
             falseNegatives += 1
148
       performanceFile = open(join(results_dir_weights
149
     'performance' + modelNumber + '.txt'), 'w')
150
       performanceFile.write(
   '************************************/n')
151
       performanceFile.write('confusion matrix: \n')
```

```
performanceFile.write('true positives: {}\n'.
152
   format(truePositives))
      performanceFile.write('false positives: {}\n'.
153
   format(falsePositives))
154
      performanceFile.write('true negatives: {}\n'.
   format(trueNegatives))
      performanceFile.write('false negatives: {}\n'.
155
   format(falseNegatives))
      performanceFile.write(
156
   157
      modelPrecision = (truePositives) / (
158
   truePositives + falsePositives)
      modelRecall = (truePositives) / (truePositives
159
    + falseNegatives)
      modelF1score = (2 * (modelPrecision *
160
   modelRecall)) / (modelPrecision + modelRecall)
161
162
      performanceFile.write(
   '************************************/n')
      performanceFile.write('Performance metrics: \n')
163
      performanceFile.write('Model Precision: {}\n'.
164
   format(modelPrecision))
      performanceFile.write('Model Recall: {}\n'.
165
   format(modelRecall))
      performanceFile.write('Model F1 Score: {}\n'.
166
   format(modelF1score))
167
      performanceFile.write(
   168
169
      performanceFile.close()
170
171
   172
   173
             = history.history['accuracy']
174
      acc
      val_acc = history.history['val_accuracy']
175
             = history.history['loss']
176
      val_loss = history.history['val_loss']
177
178
      epochs = range(len(acc)) #Get number of epochs
```

```
File - G:\GIEyA\TFG\meteor_classification\cnnTraining.py
179
         plt.plot(epochs, acc)
180
181
         plt.plot(epochs, val_acc)
         plt.title('Meteor detection training and
182
    validation accuracy')
183
184
         plt.figure()
         plt.plot(epochs, loss)
185
         plt.plot(epochs, val_loss)
186
         plt.title('Meteor detection training and
187
    validation loss')
188
         plt.show()
189
190
191
192 if __name__ == '__main__':
         p = multiprocessing.Process(target=trainCNN)
193
194
         p.start()
         p.join()
195
196
197
```