```
1 import tensorflow as tf
 2 from tensorflow.keras.callbacks import
   LearningRateScheduler
 3 from tensorflow.keras.layers import Dense, Conv2D,
   MaxPooling2D, Dropout, Flatten
 4 from tensorflow.keras.optimizers import Adam, SGD,
   RMSprop
 5 from tensorflow.keras.preprocessing.image import
   ImageDataGenerator
 6 import numpy as np
 7 import matplotlib.pyplot as plt
8 import os
9 import multiprocessing
10
11 def trainCNN( ):
12
13
       tf.keras.backend.clear_session()
14
15
       base_dir = 'G:\GIEyA\TFG\meteor_classification\
   labeledData\evenData'
16
       train_dir = os.path.join(base_dir, 'train')
17
       validation_dir = os.path.join(base_dir, 'valid')
18
19
       train_meteors_dir = os.path.join(train_dir, '
   meteors')
20
       train_non_meteors_dir = os.path.join(train_dir, '
   non_meteors')
21
       validation_meteors_dir = os.path.join(
   validation_dir, 'meteors')
22
       validation_non_meteors_dir = os.path.join(
   validation_dir, 'non_meteors')
23
24
       print('total training meteors images: ', len(os.
   listdir(train_meteors_dir)))
25
       print('total training non-meteors images: ', len(
   os.listdir(train_non_meteors_dir)))
26
       print('total validation meteors images: ', len(os
   .listdir(validation_meteors_dir)))
27
       print('total validation non-meteors images: ',
   len(os.listdir(validation_non_meteors_dir)))
28
29
30
       #Rescale all images by 1./255
31
```

```
32
       train_datagen = ImageDataGenerator(rescale=1.0/
   255,
33
                                            rotation_range
   =10, # Range from 0 to 180 degrees to randomly rotate
    images
34
   width_shift_range=0.05,
35
   height_shift_range=0.05,
36
                                            shear_range=5
    # Shear the image by 5 degrees
37
                                            zoom_range=0.1
38
   horizontal_flip=True,
39
                                            vertical_flip=
   True,
40
                                            fill_mode='
   nearest'
41
                                            )
42
43
       test_datagen = ImageDataGenerator(rescale=1.0/255
   .)
44
45
       train_generator = train_datagen.
   flow_from_directory(train_dir,
46
      batch_size=48,
47
      class_mode='binary',
48
      color_mode='grayscale',
49
      target_size=(640, 360))
50
       validation_generator = test_datagen.
   flow_from_directory(validation_dir,
51
          batch_size=48,
52
          class_mode='binary',
53
          color_mode='grayscale',
54
          target_size=(640, 360))
```

```
55
56
57
       model = tf.keras.models.Sequential([
           Conv2D(16, (2,2), activation='relu',
58
   input_shape=(640, 360, 1)),
           MaxPooling2D(2,2),
59
60
           Dropout(0.2),
           Conv2D(16, (2, 2), activation='relu'),
61
           MaxPooling2D(2, 2),
62
           #Dropout(0.2), # No dropout in transitions
63
           Conv2D(12, (2, 2), activation='relu'),
64
65
           MaxPooling2D(2, 2),
           Dropout(0.2),
66
           Conv2D(12, (2, 2), activation='relu'),
67
           #MaxPooling2D(2, 2),
68
           #Dropout(0.2), # No dropout in transitions
69
           Conv2D(12, (2, 2), activation='relu'),
70
71
           MaxPooling2D(2, 2),
           Dropout(0.2),
72
           Conv2D(8, (2, 2), activation='relu'),
73
           MaxPooling2D(2, 2),
74
75
           Dropout(0.2),
76
           Conv2D(8, (2, 2), activation='relu'),
           Dropout(0.2),
77
           Conv2D(8, (2, 2), activation='relu'),
78
           Dropout(0.2),
79
           Conv2D(4, (2, 2), activation='relu'),
80
81
           Dropout(0.2),
           Conv2D(4, (2, 2), activation='relu'),
82
83
           Dropout(0.2),
           Conv2D(4, (2, 2), activation='relu'),
84
85
           MaxPooling2D(2, 2),
           #Dropout(0.2), # No dropout in transitions
86
87
           Flatten(),
           Dense(48, activation='relu'),
88
           Dense(24, activation='relu'),
89
           Dense(12, activation='relu'),
90
91
           Dense(1, activation='sigmoid')])
92
93
       print(model.summary())
       optimizer = Adam(learning_rate=3e-3)
94
95
       model.compile(optimizer=optimizer,
96
                      loss='binary_crossentropy',
97
                      metrics=['accuracy'])
```

```
98
 99
         \#39.480 \rightarrow \text{Training } 39480 = 2 \times 2 \times 2 \times 3 \times 5 \times 6
    7 \times 47
        \#9.872 \rightarrow Validation = 2 \times 2 \times 2 \times 2 \times 617
100
101
102
         history = model.fit(train_generator,
103
                               validation_data=
    validation_generator,
104
                               steps_per_epoch=822,
105
                               epochs=10, #Later train with
     more epochs if neccessary
106
                               validation_steps=205,
107
                               verbose=1)
108
109
                   = history.history['accuracy']
         acc
         val_acc = history.history['val_accuracy']
110
                   = history.history['loss']
111
         loss
112
         val_loss = history.history['val_loss']
         epochs = range(len(acc)) #Get number of epochs
113
114
         plt.plot(epochs, acc)
115
         plt.plot(epochs, val_acc)
116
         plt.title('Meteor detection training and
117
    validation accuracy')
118
119
         plt.figure()
         plt.plot(epochs, loss)
120
         plt.plot(epochs, val_loss)
121
         plt.title('Meteor detection training and
122
    validation loss')
123
124
         plt.show()
125
126 if __name__ == '__main__':
         p = multiprocessing.Process(target=trainCNN)
127
128
         p.start()
129
         p.join()
```