```
#Load tensorflow
    import tensorflow as tf
    import glob
    #create excel data
6
    import pandas as pd1
7
    import pandas as pd2
    df train = pd1.DataFrame(columns=['class','dir','image file'])
8
9
    df test = pd2.DataFrame(columns=['class','dir','image file'])
10
11
    train cats dogs=list()
12
   train cats dogs class=list()
13
    test cats dogs=list()
14
    test cats dogs class=list()
15
16
    #Parse .jpg image files located in /data/*/*/ppg and make train, test csv.
17
18
    for entry in glob.glob('./data/*/*/pg'):
19
        cats dogs = entry
20
        if "/train" in cats dogs:
21
            train cats dogs.append(cats dogs)
22
            cats dogs dir = cats dogs[0:cats dogs.rfind('/')]
23
            cats dogs class = cats dogs dir[cats dogs dir.rfind('/')+1:]
24
            df_train.loc[df_train.shape[0]] = [cats_dogs_class, cats_dogs_dir, cats_dogs]
25
26
        if "/test" in cats dogs:
27
28
            test_cats_dogs.append(cats_dogs)
29
            cats dogs dir = cats dogs[0:cats dogs.rfind('/')]
30
            cats dogs class = cats dogs dir[cats dogs dir.rfind('/')+1:]
31
            df test.loc[df test.shape[0]] = [cats dogs class, cats dogs dir, cats dogs]
32
33
34
    print('Total images: ', df train.shape[0])
35
    df train.head()
36
37
    print('Total images: ', df test.shape[0])
38
    df test.head()
39
40 Total images: 40
41 Total images: 20
42 class dir image file
43 0 cats ./data/test/cats
                                   ./data/test/cats/110.jpg
44 1 cats
               ./data/test/cats
                                    ./data/test/cats/102.jpg
45
    2 cats ./data/test/cats
                                    ./data/test/cats/107.jpg
               ./data/test/cats
                                   ./data/test/cats/105.jpg
46
    3 cats
47
    4 cats
                ./data/test/cats
                                   ./data/test/cats/109.jpg
48
49
    #images by class
50
   print(df_train.groupby(['class']).size())
51
    print(df test.groupby(['class']).size())
52
53
   #create .csv files
54
   df train.to csv('data/train.csv',index=False)
55
   df test.to csv('data/test.csv',index=False)
56
57
   class
58 cats
            20
59
   dogs
            20
60
   dtype: int64
61 class
62 cats
            10
63 dogs
           10
64 dtype: int64
65
66 #Get class names
67 class_names = df_train['class'].unique().tolist()
    print('Animal classes: ', class_names)
68
69
    Animal classes: ['cats', 'dogs']
```

```
70
      #Define some parameters
 71
      img size = 60
 72
      img depth = 3
 73
 74
 75
      #Build batch generator
 76
      import numpy as np
 77
      def batch generator(df, batchsize=32, train mode=True):
 78
 79
          img generator= tf.keras.preprocessing.image.ImageDataGenerator(rotation range=20,
 80
                                                                            width shift range=0.2,
 81
                                                                            height shift range=0.2
 82
                                                                            horizontal flip=True,
 83
                                                                            featurewise center=Tru
                                                                            e)
 84
 85
 86
          while True:
 87
 88
                              img_class = df.loc[batch_nums[i]]['class']
 89
              #Generate random numbers to pick images from dataset
 90
              batch nums = np.random.randint(0,df.shape[0],batchsize)
 91
 92
              #Initialize batch images array
              batch images = np.zeros((batchsize,img_size, img_size,img_depth))
 93
 94
 9.5
              #Initiate batch label array
 96
              batch labels = np.zeros((batchsize, len(class names)))
 97
 98
              for i in range (batchsize):
 99
100
                   #Load image
101
                   animal image = tf.keras.preprocessing.image.load_img(
102
                                                                           df.loc[batch nums[i]]['
                                                                           image_file'],
103
                                                                           target size=(img size,
                                                                           img size))
104
                   #Convert to array
105
                   animal image = tf.keras.preprocessing.image.img to array(animal image)
106
107
                   if(train mode):
108
                       #Apply transform
109
                       animal image = img generator.random transform(animal image)
110
111
112
                   #Get the class
113
                   img class = df.loc[batch nums[i]]['class']
114
                   #Conver class to number
115
                   img class = class names.index(img class)
116
                   #Convert class to one hot encoding
117
                   img_class = tf.keras.utils.to_categorical(img_class,
                   num_classes=len(class_names))
118
119
                   #Update batch images and class arrays
120
                   batch_images[i] = animal_image
121
                   batch_labels[i] = img_class
122
                   #print("I value:", i)
123
124
              yield batch images, batch labels
125
126
127
      #Build CNN Model
128
      model = tf.keras.models.Sequential()
129
      model.add(tf.keras.layers.BatchNormalization(input shape=(img size,img size,3,)))
```

```
model.add(tf.keras.layers.Conv2D(32, kernel size=(5,5), strides=2,
130
    activation=tf.keras.layers.LeakyReLU(alpha=0.05)))
131
    model.add(tf.keras.layers.MaxPool2D(pool size=(2,2)))
    model.add(tf.keras.layers.Conv2D(64, kernel size=(5,5), strides=2,
132
    activation=tf.keras.layers.LeakyReLU(alpha=0.05)))
133
    model.add(tf.keras.layers.MaxPool2D(pool size=(2,2)))
134
    model.add(tf.keras.layers.Flatten())
135
    model.add(tf.keras.layers.Dense(200, activation=tf.keras.layers.LeakyReLU(alpha=0.03)))
136
    model.add(tf.keras.layers.Dropout(0.4))
137
    #model.add(tf.keras.layers.Flatten())
138
    model.add(tf.keras.layers.Dense(64, activation=tf.keras.layers.LeakyReLU(alpha=0.03)))
139
    #Output layer
140
    model.add(tf.keras.layers.Dense(len(class names), activation='sigmoid'))
    model.compile(optimizer='adam', loss='binary crossentropy', metrics=['accuracy'])
141
142
143
    #Train the model
    #Create train and test generator
144
145
    batchsize = 64
146
    train generator = batch generator(df train, batchsize=batchsize) #batchsize can be
147
    test generator = batch generator(df test, batchsize=batchsize, train mode=False)
148
149
    model.fit generator(train generator,
150
                      epochs=100,
151
                      steps per epoch=64,
152
                      validation data=test generator,
153
                      validation steps=32,
154
                      shuffle=True)
155
   WARNING:tensorflow:sample weight modes were coerced from
156
     . . .
157
      to
     ['...']
158
159
   WARNING: tensorflow: sample weight modes were coerced from
160
161
162
     ['...']
    Train for 64 steps, validate for 32 steps
163
164
    Epoch 1/100
165
    0.6382 - val loss: 0.9095 - val accuracy: 0.5562
166
    Epoch 2/100
    167
    0.8265 - val loss: 0.8330 - val accuracy: 0.7283
168
    Epoch 3/100
169
    0.9003 - val loss: 0.6090 - val accuracy: 0.8032
170
    Epoch 4/100
    171
    0.9539 - val loss: 0.9444 - val accuracy: 0.5913
172
    Epoch 5/100
173
    0.9634 - val loss: 1.3734 - val accuracy: 0.6460
174
    Epoch 6/100
    0.9767 - val loss: 1.0670 - val accuracy: 0.6333
176
    Epoch 7/100
177
    0.9902 - val loss: 1.3865 - val accuracy: 0.6958
178
    Epoch 8/100
    179
    0.9833 - val_loss: 1.3593 - val_accuracy: 0.5620
180
    Epoch 9/100
    181
    0.9905 - val loss: 1.6068 - val accuracy: 0.7314
    Epoch 10/100
182
    183
    0.9851 - val loss: 1.9976 - val_accuracy: 0.5732
184
    Epoch 11/100
    185
```

```
0.9933 - val loss: 2.4647 - val accuracy: 0.6582
186
  Epoch 12/100
187
  0.9957 - val loss: 2.6516 - val_accuracy: 0.5693
188
  Epoch 13/100
  189
  0.9894 - val loss: 2.3742 - val accuracy: 0.5679
190
  Epoch 14/100
  191
  0.9949 - val loss: 2.9684 - val accuracy: 0.5493
  Epoch 15/100
192
  193
  0.9935 - val loss: 2.9727 - val accuracy: 0.5120
194
  Epoch 16/100
195
  0.9952 - val loss: 2.7030 - val accuracy: 0.5894
196
  Epoch 17/100
  197
  0.9976 - val loss: 2.9585 - val accuracy: 0.5342
  Epoch 18/100
198
  199
  0.9954 - val loss: 3.0759 - val accuracy: 0.5952
200
  Epoch 19/100
  201
  0.9891 - val loss: 2.8800 - val accuracy: 0.6089
202
  Epoch 20/100
  203
  0.9952 - val loss: 3.8784 - val accuracy: 0.6567
204
  Epoch 21/100
  205
  0.9967 - val loss: 3.3587 - val accuracy: 0.6445
206
  Epoch 22/100
207
  0.9952 - val loss: 2.7973 - val accuracy: 0.6357
208
  Epoch 23/100
  209
  0.9913 - val loss: 3.2062 - val accuracy: 0.6514
210
  Epoch 24/100
211
  0.9884 - val loss: 2.7728 - val accuracy: 0.5386
212
  Epoch 25/100
  213
  0.9965 - val loss: 2.4382 - val accuracy: 0.5654
214
  Epoch 26/100
215
  0.9991 - val loss: 4.0387 - val accuracy: 0.5820
216
  Epoch 27/100
  217
  0.9980 - val loss: 3.4405 - val accuracy: 0.6992
218
  Epoch 28/100
219
  0.9987 - val loss: 3.7080 - val accuracy: 0.6445
220
  Epoch 29/100
  221
  0.9995 - val loss: 3.7287 - val accuracy: 0.6118
222
  Epoch 30/100
223
  0.9996 - val loss: 3.3922 - val accuracy: 0.6035
224
  Epoch 31/100
  225
  0.9989 - val_loss: 3.9576 - val_accuracy: 0.5879
226
  Epoch 32/100
  227
  0.9999 - val loss: 4.0844 - val accuracy: 0.5967
  Epoch 33/100
228
  229
  0.9963 - val loss: 2.5933 - val_accuracy: 0.6311
230
  Epoch 34/100
  231
```

```
0.9967 - val loss: 3.1865 - val accuracy: 0.6726
232
  Epoch 35/100
233
  0.9984 - val loss: 4.2086 - val_accuracy: 0.6123
234
  Epoch 36/100
  235
  0.9976 - val loss: 3.0882 - val accuracy: 0.6475
  Epoch 37/100
236
  237
  0.9907 - val loss: 2.9431 - val accuracy: 0.5967
  Epoch 38/100
238
  239
  0.9945 - val loss: 2.4969 - val accuracy: 0.7051
  Epoch 39/100
240
  241
  0.9974 - val loss: 3.6264 - val accuracy: 0.7017
242
  Epoch 40/100
  243
  0.9987 - val loss: 4.6557 - val accuracy: 0.5464
  Epoch 41/100
244
  245
  0.9996 - val loss: 4.2020 - val accuracy: 0.5972
246
  Epoch 42/100
  247
  0.9989 - val loss: 4.1630 - val accuracy: 0.5610
248
  Epoch 43/100
249
  0.9988 - val loss: 3.6876 - val accuracy: 0.5991
250
  Epoch 44/100
  2.51
  0.9994 - val loss: 4.2227 - val accuracy: 0.6484
252
  Epoch 45/100
  0.9971 - val loss: 3.3744 - val accuracy: 0.5679
254
  Epoch 46/100
  255
  0.9962 - val loss: 2.9655 - val accuracy: 0.7305
256
  Epoch 47/100
257
  0.9961 - val loss: 3.9505 - val accuracy: 0.5498
258
  Epoch 48/100
  259
  0.9958 - val loss: 5.0354 - val accuracy: 0.5562
260
  Epoch 49/100
261
  0.9943 - val loss: 4.9747 - val accuracy: 0.4541
262
  Epoch 50/100
  263
  0.9934 - val_loss: 3.1141 - val_accuracy: 0.6294
  Epoch 51/100
264
265
  0.9995 - val loss: 3.1312 - val accuracy: 0.5562
266
  Epoch 52/100
  267
  1.0000 - val loss: 3.5130 - val accuracy: 0.5483
268
  Epoch 53/100
269
  1.0000 - val loss: 4.1483 - val accuracy: 0.5898
270
  Epoch 54/100
  271
  1.0000 - val_loss: 4.5619 - val_accuracy: 0.5522
272
  Epoch 55/100
  273
  0.9998 - val loss: 4.1330 - val accuracy: 0.6978
274
  Epoch 56/100
  275
  0.9940 - val loss: 3.9637 - val_accuracy: 0.5776
276
  Epoch 57/100
  277
```

```
0.9819 - val loss: 2.7062 - val accuracy: 0.7329
278
  Epoch 58/100
279
  0.9988 - val loss: 2.9132 - val_accuracy: 0.6218
280
  Epoch 59/100
  281
  0.9987 - val loss: 2.8939 - val accuracy: 0.6353
  Epoch 60/100
282
  283
  1.0000 - val loss: 3.6556 - val accuracy: 0.5864
  Epoch 61/100
284
  285
  0.9998 - val loss: 3.2075 - val accuracy: 0.6499
  Epoch 62/100
286
  287
  0.9993 - val loss: 3.2318 - val accuracy: 0.6562
288
  Epoch 63/100
  289
  0.9993 - val loss: 3.8935 - val accuracy: 0.6069
  Epoch 64/100
290
  291
  0.9968 - val loss: 2.8898 - val accuracy: 0.6416
292
  Epoch 65/100
293
  0.9991 - val loss: 2.8350 - val accuracy: 0.6426
294
  Epoch 66/100
  295
  0.9983 - val loss: 3.2272 - val accuracy: 0.5977
296
  Epoch 67/100
  297
  0.9995 - val loss: 3.9669 - val accuracy: 0.5967
298
  Epoch 68/100
299
  0.9995 - val loss: 3.8911 - val accuracy: 0.5552
300
  Epoch 69/100
  301
  1.0000 - val loss: 3.8883 - val accuracy: 0.6045
302
  Epoch 70/100
303
  1.0000 - val loss: 4.0434 - val accuracy: 0.5649
304
  Epoch 71/100
  305
  1.0000 - val loss: 3.8027 - val accuracy: 0.5986
306
  Epoch 72/100
307
  1.0000 - val loss: 2.8565 - val accuracy: 0.6147
308
  Epoch 73/100
  309
  0.9989 - val loss: 3.6815 - val accuracy: 0.6301
  Epoch 74/100
310
311
  0.9969 - val loss: 2.9839 - val accuracy: 0.6621
312
  Epoch 75/100
  313
  0.9979 - val loss: 3.3570 - val accuracy: 0.4951
314
  Epoch 76/100
315
  1.0000 - val loss: 4.5387 - val accuracy: 0.5044
316
  Epoch 77/100
  317
  0.9999 - val_loss: 3.2457 - val_accuracy: 0.5774
318
  Epoch 78/100
  319
  0.9984 - val loss: 4.0395 - val accuracy: 0.5518
  Epoch 79/100
320
  321
  0.9961 - val loss: 3.6527 - val_accuracy: 0.6245
322
  Epoch 80/100
  323
```

```
0.9952 - val loss: 3.2264 - val accuracy: 0.6084
324
  Epoch 81/100
325
  0.9984 - val loss: 2.8905 - val accuracy: 0.6245
326
  Epoch 82/100
  327
  0.9940 - val loss: 1.5342 - val accuracy: 0.6626
  Epoch 83/100
328
329
  0.9995 - val loss: 3.9380 - val accuracy: 0.6257
330
  Epoch 84/100
  331
  0.9982 - val loss: 5.4605 - val accuracy: 0.5557
332
  Epoch 85/100
  333
  0.9978 - val loss: 5.9055 - val accuracy: 0.6460
334
  Epoch 86/100
  335
  0.9983 - val loss: 3.4676 - val accuracy: 0.6064
336
  Epoch 87/100
  337
  0.9988 - val loss: 4.5562 - val accuracy: 0.6016
338
  Epoch 88/100
  339
  1.0000 - val loss: 4.7289 - val accuracy: 0.6284
340
  Epoch 89/100
  341
  0.9996 - val loss: 5.0640 - val accuracy: 0.6128
342
  Epoch 90/100
  343
  1.0000 - val loss: 4.9349 - val accuracy: 0.6064
344
  Epoch 91/100
  1.0000 - val loss: 5.4669 - val accuracy: 0.5923
346
  Epoch 92/100
  347
  1.0000 - val loss: 4.8902 - val accuracy: 0.5562
348
  Epoch 93/100
349
  0.9987 - val loss: 4.0293 - val accuracy: 0.5649
350
  Epoch 94/100
  351
  0.9985 - val loss: 2.7171 - val accuracy: 0.5532
352
  Epoch 95/100
353
  0.9987 - val loss: 4.3506 - val accuracy: 0.6382
354
  Epoch 96/100
  355
  0.9962 - val loss: 6.4583 - val accuracy: 0.5149
356
  Epoch 97/100
357
  0.9957 - val loss: 3.3490 - val accuracy: 0.6118
  Epoch 98/100
358
  359
  0.9994 - val loss: 5.1457 - val accuracy: 0.5537
360
  Epoch 99/100
  361
  0.9990 - val loss: 4.5643 - val accuracy: 0.6035
362
  Epoch 100/100
  363
  1.0000 - val loss: 4.7228 - val_accuracy: 0.5869
364
  <tensorflow.python.keras.callbacks.History at 0x7f530c47d4d0>
```

365 366