CNNModel

April 17, 2020

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[2]: import keras
      import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
      from keras.models import Sequential
      from keras.applications.vgg16 import VGG16
      from keras.layers import Dense, InputLayer, Dropout, Flatten
      from keras.layers import Conv2D, MaxPooling2D, GlobalMaxPooling2D
      from keras.preprocessing import image
      from sklearn.model_selection import train_test_split
      from tqdm import tqdm
 [5]: data = pd.read_csv('dataset.csv')
      data.head()
 [5]:
                                 image label
      0
             Dataset/2_241_gel.wav.png
                                         gel
             Dataset/2_580_voi.wav.png
      1
                                         voi
      2 Dataset/0_214_pia_sax.wav.png
                                         pia
      3
             Dataset/0_198_voi.wav.png
                                         voi
      4 Dataset/0_144_gel_voi.wav.png
                                         gel
[15]: train_im = data['image'][:1200]
      test_im = data['image'][1200:]
      train_y = data['label'][:1200]
      test_y = data['label'][1200:]
[16]: train = pd.DataFrame()
      test = pd.DataFrame()
      train['image'] = train_im
      test['image'] = test_im
      train['label'] = train_y
      test['label'] = test_y
[40]: train.to_csv('training.csv', header=True, index = False)
      test.to_csv('testing.csv', header = True, index = False)
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[43]: train_images = []
      for i in tqdm(range(1200)):
          im = image.load_img(train['image'][i], target_size = (224, 224))
          im = image.img_to_array(im)
          im = im/255
          train_images.append(im)
     100%|
                | 1200/1200 [00:02<00:00, 561.49it/s]
[44]: X = np.array(train_images)
[45]: y = train['label']
[46]: y.shape
[46]: (1200,)
[47]: X_train, X_test,y_train, y_test = train_test_split(X, y, random_state = 42,__
       →test_size = 0.2, stratify = y)
[48]: X.shape
[48]: (1200, 224, 224, 3)
[49]: y.shape
[49]: (1200,)
[50]: y_train = pd.get_dummies(y_train)
      y_test = pd.get_dummies(y_test)
[51]: y_train.shape
[51]: (960, 11)
[52]: y_train
[52]:
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496
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     1047
     [960 rows x 11 columns]
[53]: basemodel = VGG16(weights = 'imagenet', include_top = False)
[54]: X_train = basemodel.predict(X_train)
[55]: X_train.shape
[55]: (960, 7, 7, 512)
[56]: X_test = basemodel.predict(X_test)
[57]: X_test.shape
[57]: (240, 7, 7, 512)
[58]: X_train = X_train.reshape(960, 7*7*512)
     X_{\text{test}} = X_{\text{test.reshape}}(240, 7*7*512)
[59]: maxi = X_train.max()
     X_train = X_train/maxi
     X_{test} = X_{test/maxi}
[60]: X_train.shape
[60]: (960, 25088)
[61]: model = Sequential()
     model.add(Dense(1024, activation='relu', input_shape=(25088,)))
     model.add(Dropout(0.5))
     model.add(Dense(512, activation='relu'))
     model.add(Dropout(0.5))
     model.add(Dense(256, activation='relu'))
     model.add(Dropout(0.5))
     model.add(Dense(128, activation='relu'))
     model.add(Dropout(0.5))
     model.add(Dense(11, activation='softmax'))
[62]: from keras.callbacks import ModelCheckpoint
     mcp_save = ModelCheckpoint('weight.hdf5', save_best_only=True,__
      →monitor='val_loss', mode='min')
[63]: model.
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Train on 960 samples, validate on 240 samples
Epoch 1/100
960/960 [============ ] - 1s 1ms/step - loss: 2.2031 -
accuracy: 0.2427 - val_loss: 1.9809 - val_accuracy: 0.3292
Epoch 2/100
960/960 [============ ] - 1s 1ms/step - loss: 2.0180 -
accuracy: 0.2958 - val_loss: 2.0382 - val_accuracy: 0.2500
Epoch 3/100
960/960 [============ ] - 1s 1ms/step - loss: 2.0087 -
accuracy: 0.2750 - val_loss: 1.9922 - val_accuracy: 0.3292
Epoch 4/100
960/960 [============ ] - 1s 1ms/step - loss: 1.9429 -
accuracy: 0.2917 - val_loss: 1.9753 - val_accuracy: 0.4250
Epoch 5/100
accuracy: 0.2823 - val_loss: 1.9279 - val_accuracy: 0.4083
Epoch 6/100
960/960 [============ ] - 1s 1ms/step - loss: 1.8890 -
accuracy: 0.3094 - val_loss: 1.9000 - val_accuracy: 0.4292
Epoch 7/100
accuracy: 0.3063 - val_loss: 1.9543 - val_accuracy: 0.3833
Epoch 8/100
accuracy: 0.3167 - val_loss: 1.9287 - val_accuracy: 0.3500
Epoch 9/100
accuracy: 0.3104 - val_loss: 1.8568 - val_accuracy: 0.4167
960/960 [============ ] - 1s 1ms/step - loss: 1.8471 -
accuracy: 0.3490 - val_loss: 1.8403 - val_accuracy: 0.4167
Epoch 11/100
960/960 [============ ] - 1s 1ms/step - loss: 1.8390 -
accuracy: 0.3167 - val_loss: 1.7784 - val_accuracy: 0.4542
Epoch 12/100
960/960 [=========== ] - 1s 1ms/step - loss: 1.8225 -
accuracy: 0.3302 - val_loss: 1.8327 - val_accuracy: 0.3958
Epoch 13/100
accuracy: 0.3469 - val_loss: 1.7609 - val_accuracy: 0.4458
Epoch 14/100
accuracy: 0.3552 - val_loss: 1.7339 - val_accuracy: 0.4708
Epoch 15/100
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accuracy: 0.3688 - val_loss: 1.7268 - val_accuracy: 0.4542
Epoch 16/100
accuracy: 0.4052 - val loss: 1.6554 - val accuracy: 0.4667
Epoch 17/100
accuracy: 0.4073 - val_loss: 1.6029 - val_accuracy: 0.4917
Epoch 18/100
accuracy: 0.4302 - val_loss: 1.6088 - val_accuracy: 0.4875
Epoch 19/100
accuracy: 0.4177 - val_loss: 1.5595 - val_accuracy: 0.4667
Epoch 20/100
accuracy: 0.4479 - val_loss: 1.5179 - val_accuracy: 0.4917
Epoch 21/100
accuracy: 0.4646 - val_loss: 1.5228 - val_accuracy: 0.5167
Epoch 22/100
accuracy: 0.4625 - val_loss: 1.5203 - val_accuracy: 0.4750
Epoch 23/100
accuracy: 0.4542 - val_loss: 1.4982 - val_accuracy: 0.5208
Epoch 24/100
960/960 [============ ] - 1s 1ms/step - loss: 1.5211 -
accuracy: 0.4604 - val_loss: 1.4883 - val_accuracy: 0.5208
Epoch 25/100
960/960 [============ ] - 1s 1ms/step - loss: 1.5256 -
accuracy: 0.4854 - val_loss: 1.4316 - val_accuracy: 0.5292
Epoch 26/100
960/960 [============ ] - 1s 1ms/step - loss: 1.4475 -
accuracy: 0.4844 - val loss: 1.4023 - val accuracy: 0.5042
Epoch 27/100
960/960 [============ ] - 1s 1ms/step - loss: 1.4323 -
accuracy: 0.5125 - val_loss: 1.3929 - val_accuracy: 0.5708
Epoch 28/100
accuracy: 0.5052 - val_loss: 1.3882 - val_accuracy: 0.5625
Epoch 29/100
960/960 [============ ] - 1s 1ms/step - loss: 1.3780 -
accuracy: 0.5344 - val_loss: 1.3809 - val_accuracy: 0.5333
Epoch 30/100
960/960 [============ ] - 1s 1ms/step - loss: 1.3357 -
accuracy: 0.5365 - val_loss: 1.3627 - val_accuracy: 0.5500
Epoch 31/100
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accuracy: 0.5406 - val_loss: 1.3642 - val_accuracy: 0.5250
Epoch 32/100
accuracy: 0.5500 - val loss: 1.3376 - val accuracy: 0.5625
Epoch 33/100
accuracy: 0.5417 - val_loss: 1.3963 - val_accuracy: 0.5500
Epoch 34/100
960/960 [============ ] - 1s 1ms/step - loss: 1.2813 -
accuracy: 0.5698 - val_loss: 1.3553 - val_accuracy: 0.5667
Epoch 35/100
accuracy: 0.5469 - val_loss: 1.3780 - val_accuracy: 0.5708
Epoch 36/100
accuracy: 0.5646 - val_loss: 1.3634 - val_accuracy: 0.5667
Epoch 37/100
accuracy: 0.5604 - val_loss: 1.3524 - val_accuracy: 0.5708
Epoch 38/100
960/960 [============ ] - 1s 1ms/step - loss: 1.1993 -
accuracy: 0.5917 - val_loss: 1.2886 - val_accuracy: 0.6000
Epoch 39/100
accuracy: 0.6000 - val_loss: 1.2948 - val_accuracy: 0.5708
Epoch 40/100
960/960 [=========== ] - 1s 1ms/step - loss: 1.1497 -
accuracy: 0.6156 - val_loss: 1.3092 - val_accuracy: 0.5708
Epoch 41/100
960/960 [============ ] - 1s 1ms/step - loss: 1.1723 -
accuracy: 0.5917 - val_loss: 1.4022 - val_accuracy: 0.5750
Epoch 42/100
960/960 [============ ] - 1s 1ms/step - loss: 1.1293 -
accuracy: 0.6073 - val_loss: 1.3199 - val_accuracy: 0.5750
Epoch 43/100
960/960 [============ ] - 1s 1ms/step - loss: 1.0786 -
accuracy: 0.6250 - val_loss: 1.3146 - val_accuracy: 0.5542
Epoch 44/100
accuracy: 0.6281 - val_loss: 1.4064 - val_accuracy: 0.5875
Epoch 45/100
960/960 [============ ] - 1s 1ms/step - loss: 1.0643 -
accuracy: 0.6302 - val_loss: 1.3798 - val_accuracy: 0.5875
Epoch 46/100
960/960 [============ ] - 1s 1ms/step - loss: 1.0417 -
accuracy: 0.6479 - val_loss: 1.3789 - val_accuracy: 0.5708
Epoch 47/100
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accuracy: 0.6625 - val_loss: 1.4242 - val_accuracy: 0.5542
Epoch 48/100
accuracy: 0.6115 - val loss: 1.3829 - val accuracy: 0.5958
Epoch 49/100
accuracy: 0.6479 - val_loss: 1.5316 - val_accuracy: 0.5500
Epoch 50/100
960/960 [============ ] - 1s 1ms/step - loss: 1.0740 -
accuracy: 0.6417 - val_loss: 1.3169 - val_accuracy: 0.6042
Epoch 51/100
accuracy: 0.6615 - val_loss: 1.3238 - val_accuracy: 0.6167
Epoch 52/100
accuracy: 0.6604 - val_loss: 1.3487 - val_accuracy: 0.6042
Epoch 53/100
accuracy: 0.6438 - val_loss: 1.3820 - val_accuracy: 0.5833
Epoch 54/100
960/960 [============ ] - 1s 1ms/step - loss: 1.0559 -
accuracy: 0.6271 - val_loss: 1.3262 - val_accuracy: 0.6333
Epoch 55/100
accuracy: 0.6562 - val_loss: 1.3553 - val_accuracy: 0.6292
Epoch 56/100
960/960 [============ ] - 1s 1ms/step - loss: 1.0207 -
accuracy: 0.6562 - val_loss: 1.3364 - val_accuracy: 0.6083
Epoch 57/100
960/960 [=========== ] - 1s 1ms/step - loss: 1.0137 -
accuracy: 0.6396 - val_loss: 1.2971 - val_accuracy: 0.5917
Epoch 58/100
accuracy: 0.6562 - val_loss: 1.3419 - val_accuracy: 0.6250
Epoch 59/100
960/960 [============ ] - 1s 1ms/step - loss: 0.9812 -
accuracy: 0.6792 - val_loss: 1.3371 - val_accuracy: 0.6208
Epoch 60/100
accuracy: 0.6302 - val_loss: 1.4752 - val_accuracy: 0.5833
Epoch 61/100
960/960 [============ ] - 1s 1ms/step - loss: 0.9973 -
accuracy: 0.6625 - val_loss: 1.4036 - val_accuracy: 0.6042
Epoch 62/100
960/960 [============ ] - 1s 1ms/step - loss: 1.0573 -
accuracy: 0.6375 - val_loss: 1.2617 - val_accuracy: 0.6417
Epoch 63/100
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accuracy: 0.6167 - val_loss: 1.3994 - val_accuracy: 0.6083
Epoch 64/100
accuracy: 0.6510 - val_loss: 1.4041 - val_accuracy: 0.5667
Epoch 65/100
accuracy: 0.6094 - val_loss: 1.2841 - val_accuracy: 0.6208
Epoch 66/100
accuracy: 0.6313 - val_loss: 1.3173 - val_accuracy: 0.6125
Epoch 67/100
accuracy: 0.6156 - val_loss: 1.2972 - val_accuracy: 0.6292
Epoch 68/100
accuracy: 0.5990 - val_loss: 1.3085 - val_accuracy: 0.6250
Epoch 69/100
accuracy: 0.6094 - val_loss: 1.3238 - val_accuracy: 0.6542
Epoch 70/100
accuracy: 0.6271 - val_loss: 1.3075 - val_accuracy: 0.6250
Epoch 71/100
accuracy: 0.6365 - val_loss: 1.3250 - val_accuracy: 0.6167
Epoch 72/100
960/960 [============ ] - 1s 2ms/step - loss: 0.9975 -
accuracy: 0.6385 - val_loss: 1.3103 - val_accuracy: 0.6292
Epoch 73/100
960/960 [============ ] - 1s 1ms/step - loss: 0.9623 -
accuracy: 0.6458 - val_loss: 1.3034 - val_accuracy: 0.6250
Epoch 74/100
960/960 [============ ] - 1s 2ms/step - loss: 0.9871 -
accuracy: 0.6229 - val_loss: 1.3571 - val_accuracy: 0.6292
Epoch 75/100
960/960 [============ ] - 1s 1ms/step - loss: 0.9508 -
accuracy: 0.6542 - val_loss: 1.3726 - val_accuracy: 0.6333
Epoch 76/100
accuracy: 0.6438 - val_loss: 1.4146 - val_accuracy: 0.6250
Epoch 77/100
960/960 [============ ] - 1s 1ms/step - loss: 0.9300 -
accuracy: 0.6708 - val_loss: 1.3932 - val_accuracy: 0.6125
Epoch 78/100
960/960 [============ ] - 1s 1ms/step - loss: 0.9194 -
accuracy: 0.6635 - val_loss: 1.2851 - val_accuracy: 0.6208
Epoch 79/100
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accuracy: 0.6708 - val_loss: 1.3990 - val_accuracy: 0.6458
Epoch 80/100
accuracy: 0.6656 - val loss: 1.3853 - val accuracy: 0.6458
Epoch 81/100
accuracy: 0.6708 - val_loss: 1.4289 - val_accuracy: 0.6292
Epoch 82/100
960/960 [============ ] - 1s 1ms/step - loss: 0.8891 -
accuracy: 0.6656 - val_loss: 1.3815 - val_accuracy: 0.6042
Epoch 83/100
accuracy: 0.6687 - val_loss: 1.3831 - val_accuracy: 0.6375
Epoch 84/100
accuracy: 0.6344 - val_loss: 1.3902 - val_accuracy: 0.6083
Epoch 85/100
accuracy: 0.6792 - val_loss: 1.3690 - val_accuracy: 0.6208
Epoch 86/100
accuracy: 0.6573 - val_loss: 1.5380 - val_accuracy: 0.6458
Epoch 87/100
accuracy: 0.6708 - val_loss: 1.4371 - val_accuracy: 0.5917
Epoch 88/100
960/960 [============ ] - 1s 1ms/step - loss: 0.8797 -
accuracy: 0.6552 - val_loss: 1.6749 - val_accuracy: 0.6333
Epoch 89/100
960/960 [============ ] - 1s 1ms/step - loss: 0.8695 -
accuracy: 0.6823 - val_loss: 1.4663 - val_accuracy: 0.6167
Epoch 90/100
accuracy: 0.6521 - val_loss: 1.4745 - val_accuracy: 0.5917
Epoch 91/100
960/960 [============ ] - 1s 1ms/step - loss: 0.9445 -
accuracy: 0.6458 - val_loss: 1.5025 - val_accuracy: 0.6125
Epoch 92/100
accuracy: 0.6635 - val_loss: 1.4681 - val_accuracy: 0.6083
Epoch 93/100
960/960 [============ ] - 1s 1ms/step - loss: 0.9050 -
accuracy: 0.6562 - val_loss: 1.7391 - val_accuracy: 0.6167
Epoch 94/100
960/960 [============ ] - 1s 1ms/step - loss: 1.0169 -
accuracy: 0.6167 - val_loss: 1.3776 - val_accuracy: 0.6167
Epoch 95/100
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960/960 [=========== ] - 1s 1ms/step - loss: 0.9377 -
    accuracy: 0.6396 - val_loss: 1.5053 - val_accuracy: 0.6083
    Epoch 96/100
    960/960 [============= ] - 1s 1ms/step - loss: 0.9059 -
    accuracy: 0.6531 - val_loss: 1.4963 - val_accuracy: 0.6167
    Epoch 97/100
    accuracy: 0.6438 - val_loss: 1.4210 - val_accuracy: 0.6417
    Epoch 98/100
    960/960 [============ ] - 1s 1ms/step - loss: 0.9626 -
    accuracy: 0.6354 - val_loss: 1.4992 - val_accuracy: 0.6208
    Epoch 99/100
    960/960 [============ ] - 1s 1ms/step - loss: 1.0057 -
    accuracy: 0.6271 - val_loss: 1.4605 - val_accuracy: 0.5667
    accuracy: 0.6156 - val_loss: 1.5149 - val_accuracy: 0.5875
[64]: <keras.callbacks.dallbacks.History at 0x7f90f62a5ac8>
[]:
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