vgg16

April 2, 2025

1 Import Libraries

```
import numpy as np # linear algebra
import os
from tqdm import tqdm
from tensorflow import keras
import cv2
from sklearn.utils import shuffle
from keras.utils import to_categorical
from sklearn.model_selection import train_test_split
from sklearn.metrics import classification_report, confusion_matrix,u

GConfusionMatrixDisplay
import tensorflow
from keras.applications.vgg16 import VGG16
from keras.layers import Dense, Flatten
from keras.models import Model
import matplotlib.pyplot as plt
```

2 Load and Pre-Process data

```
image_data = np.array(image_data)
     label_data = np.array(label_data)
    pre-processing data
    100%|
              | 1341/1341 [00:29<00:00, 44.85it/s]
    100%|
              | 3875/3875 [00:37<00:00, 102.31it/s]
    100%|
              | 8/8 [00:00<00:00, 49.41it/s]
              | 8/8 [00:00<00:00, 96.62it/s]
    100%|
              | 234/234 [00:04<00:00, 57.61it/s]
    100%|
    100%|
              | 390/390 [00:04<00:00, 91.31it/s]
[4]: image_data,label_data=shuffle(image_data,label_data,random_state=42)
     X_train, X_test, Y_train, Y_test=train_test_split(image_data,label_data,test_size=0.
      →2,random_state=42)
[5]: train_label_data_new=[]
     test_label_data_new=[]
     for n in Y train:
         train_label_data_new.append(labels.index(n))
     Y_train=train_label_data_new
     Y_train=to_categorical(Y_train)
     for n in Y test:
         test_label_data_new.append(labels.index(n))
     Y test=test label data new
     Y_test=to_categorical(Y_test)
    3 VGG16 Model
[7]: vgg = VGG16(input_shape=(150,150,3), weights='imagenet', include_top=False)
     for layer in vgg.layers:
         layer.trainable = False
     x = Flatten()(vgg.output)
     prediction = Dense(2, activation='softmax')(x)
     modelvgg = Model(inputs=vgg.input, outputs=prediction)
[8]: modelvgg.compile(optimizer='adam',
     loss=tensorflow.losses.CategoricalCrossentropy(),
     metrics=[keras.metrics.AUC(name='auc')])
     callback = keras.callbacks.
```

validation_data=(X_test,Y_test), callbacks=callback)

```
Epoch 2/10
     469/469
                         18s 38ms/step -
     auc: 0.9689 - loss: 0.8028 - val_auc: 0.9573 - val_loss: 1.2216
     Epoch 3/10
     469/469
                         19s 39ms/step -
     auc: 0.9719 - loss: 0.6833 - val_auc: 0.9397 - val_loss: 2.7039
     Epoch 4/10
                         20s 42ms/step -
     469/469
     auc: 0.9820 - loss: 0.4660 - val_auc: 0.9515 - val_loss: 1.9728
     Epoch 5/10
     469/469
                         21s 44ms/step -
     auc: 0.9859 - loss: 0.2625 - val_auc: 0.9517 - val_loss: 1.8240
     Epoch 6/10
     469/469
                         23s 48ms/step -
     auc: 0.9877 - loss: 0.2921 - val_auc: 0.9484 - val_loss: 2.0615
     Epoch 7/10
     469/469
                         23s 50ms/step -
     auc: 0.9885 - loss: 0.2675 - val_auc: 0.9628 - val_loss: 1.6218
     Epoch 8/10
                         21s 46ms/step -
     auc: 0.9942 - loss: 0.1395 - val_auc: 0.9559 - val_loss: 2.6995
     Epoch 9/10
     469/469
                         22s 46ms/step -
     auc: 0.9897 - loss: 0.2813 - val_auc: 0.9518 - val_loss: 2.7672
     4 Accuracy & Loss
[10]: loss, accuracy = modelvgg.evaluate(X_train, Y_train)
      print("Train Loss: ", loss)
      print("Train Accuracy: ", accuracy)
     147/147
                         31s 131ms/step -
     auc: 0.9743 - loss: 0.4925
     Train Loss: 0.4725184440612793
     Train Accuracy: 0.9759469628334045
[11]: loss, accuracy = modelvgg.evaluate(X_test, Y_test)
      print("Test Loss: ", loss)
```

37/37 13s 351ms/step -

print("Test Accuracy: ", accuracy)

Epoch 1/10 469/469

29s 50ms/step -

auc: 0.9128 - loss: 1.6195 - val_auc: 0.9560 - val_loss: 1.0363

auc: 0.9577 - loss: 1.0455
Test Loss: 1.0363073348999023
Test Accuracy: 0.9559816718101501

```
[12]: plt.plot(history.history['auc'])
   plt.plot(history.history['val_auc'])
   plt.title('model accuracy')
   plt.ylabel('accuracy')
   plt.xlabel('epoch')
   plt.legend(['Train', 'Test'], loc='upper left')
   plt.show()
   plt.plot(history.history['loss'])
   plt.plot(history.history['val_loss'])
   plt.title('model loss')
   plt.ylabel('loss')
   plt.xlabel('epoch')
   plt.legend(['Train', 'Test'], loc='upper left')
   plt.show()
```

model accuracy



