symcnn

March 6, 2022

```
[]: import os
     import cv2
     import numpy as np
     import pandas as pd
     import seaborn as sns
     import matplotlib.pyplot as plt
     import sklearn.metrics as metrics
     from sklearn.model_selection import train_test_split
     from sklearn.preprocessing import LabelEncoder
     from sklearn.metrics import confusion_matrix, classification_report
     from sklearn.model_selection import GridSearchCV
     from sklearn import svm
     from tensorflow.python.keras.utils import np_utils
     from tensorflow.keras.models import Model
     from tensorflow.keras.models import Sequential
     from tensorflow.keras.optimizers import Adam
     from tensorflow.keras.utils import to_categorical
     from tensorflow.keras.preprocessing.image import ImageDataGenerator
     from tensorflow.keras.applications import VGG16, ResNet101, Xception
     from tensorflow.keras.layers import Input, Dense, Flatten, MaxPooling2D,
      →GlobalAveragePooling2D, Dropout, BatchNormalization, Conv2D, InputLayer
[]: SIZE = 224 #Resize images
```

```
[ ]: X = []
     y = []
     for img_path in imagePaths:
         label = img_path.split(os.path.sep)[-2]
         img = cv2.imread(img_path, cv2.IMREAD_COLOR)
         img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
         img = cv2.resize(img, (SIZE,SIZE))
         X.append(img)
         y.append(label)
     X = np.array(X)
     y = np.array(y)
     print(type(X), type(y), '\n')
     print(X.shape, y.shape)
    <class 'numpy.ndarray'> <class 'numpy.ndarray'>
    (8149, 224, 224, 3) (8149,)
[]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30,__
     →random_state=3)
[]: #Encode labels as integers
    le = LabelEncoder()
     le.fit(y_test)
     y_test_labels_encoded = le.transform(y_test)
     le.fit(y_train)
     y_train_labels_encoded = le.transform(y_train)
     type(y_train_labels_encoded)
[]: numpy.ndarray
[]: # Normalize pixel values to between 0 and 1
     X_train, X_test = X_train / 255.0, X_test / 255.0
[]: y_train_one_hot = np_utils.to_categorical(y_train_labels_encoded)
     y_test_one_hot = np_utils.to_categorical(y_test_labels_encoded)
[]: VGG_model = VGG16(weights='imagenet', include_top=False, input_shape=(SIZE,__
      \rightarrowSIZE, 3))
```

#Make loaded layers as non-trainable. This is important as we want to work with $\underline{\ }$ \rightarrow pre-trained weights

VGG_model.summary() #Trainable parameters will be 0

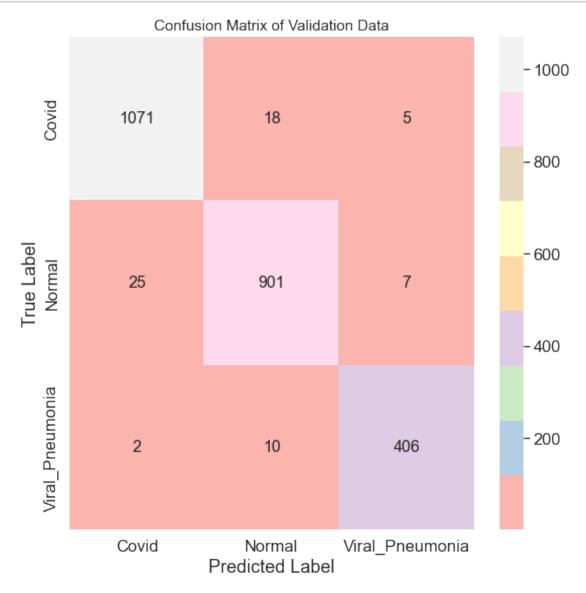
Model: "vgg16"

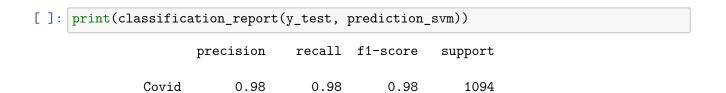
Layer (type)	Output Shape	Param #
input_1 (InputLayer)		
block1_conv1 (Conv2D)	(None, 224, 224, 64)	1792
block1_conv2 (Conv2D)	(None, 224, 224, 64)	36928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	0
block2_conv1 (Conv2D)	(None, 112, 112, 128)	73856
block2_conv2 (Conv2D)	(None, 112, 112, 128)	147584
block2_pool (MaxPooling2D)	(None, 56, 56, 128)	0
block3_conv1 (Conv2D)	(None, 56, 56, 256)	295168
block3_conv2 (Conv2D)	(None, 56, 56, 256)	590080
block3_conv3 (Conv2D)	(None, 56, 56, 256)	590080
block3_pool (MaxPooling2D)	(None, 28, 28, 256)	0
block4_conv1 (Conv2D)	(None, 28, 28, 512)	1180160
block4_conv2 (Conv2D)	(None, 28, 28, 512)	2359808
block4_conv3 (Conv2D)	(None, 28, 28, 512)	2359808
block4_pool (MaxPooling2D)	(None, 14, 14, 512)	0
block5_conv1 (Conv2D)	(None, 14, 14, 512)	2359808
block5_conv2 (Conv2D)	(None, 14, 14, 512)	2359808
block5_conv3 (Conv2D)	(None, 14, 14, 512)	2359808
block5_pool (MaxPooling2D)	(None, 7, 7, 512)	0

```
Total params: 14,714,688
    Trainable params: 0
    Non-trainable params: 14,714,688
[]: #Now, let us use features from convolutional network for RF
     feature_extractor=VGG_model.predict(X_train)
[]: features = feature_extractor.reshape(feature_extractor.shape[0], -1)
     X_for_SVM = features #This is our X input to RF
[]: # Linear Kernel
     svm_covid = svm.SVC(kernel='linear')
[]: svm_covid.fit(X_for_SVM,y_train_labels_encoded)
[]: SVC(kernel='linear')
[ ]: | X_test_feature = VGG_model.predict(X_test)
     X_test_feature = X_test_feature.reshape(X_test_feature.shape[0], -1)
[]: prediction_svm = svm_covid.predict(X_test_features)
     #Inverse le transform to get original label back.
     prediction_svm = le.inverse_transform(prediction_svm)
[]: print ("Accuracy = ", metrics.accuracy_score(y_test, prediction_svm)*100)
     cm = confusion_matrix(y_test, prediction_svm)
     print(cm)
     cm_df = pd.DataFrame(cm, index=le.classes_, columns=le.classes_)
     cm_df.head()
    Accuracy = 97.2597137014315
    [[1071
             18
                   5]
     Γ 25 901
                   71
        2
             10 406]]
[]:
                      Covid Normal Viral_Pneumonia
     Covid
                       1071
                                 18
                                                   5
                                                   7
                         25
                                901
     Normal
     Viral_Pneumonia
                          2
                                                 406
                                 10
[]: plt.figure(figsize=(9,9))
```

```
sns.set(font_scale=1.5, color_codes=True, palette='deep')
sns.heatmap(cm_df, annot=True, annot_kws={'size':16}, fmt='d', cmap='Pastel1')

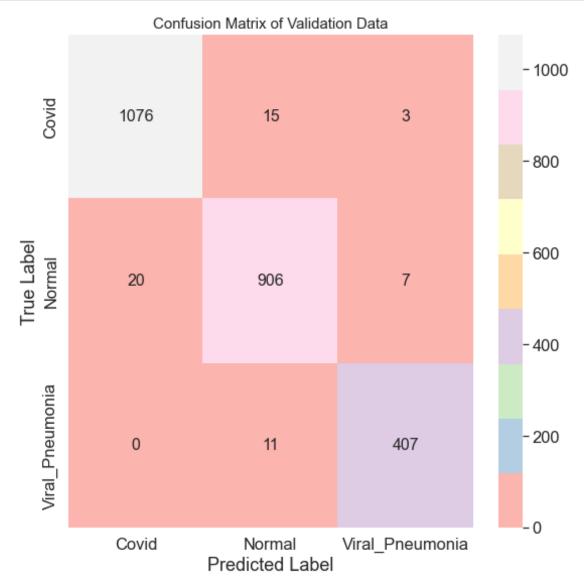
plt.ylabel("True Label")
plt.xlabel("Predicted Label")
plt.title('Confusion Matrix of Validation Data', size=15)
plt.show()
```





```
Normal
                          0.97
                                     0.97
                                               0.97
                                                          933
    Viral_Pneumonia
                           0.97
                                     0.97
                                                          418
                                               0.97
                                               0.97
                                                         2445
           accuracy
                                     0.97
                                                         2445
          macro avg
                          0.97
                                               0.97
       weighted avg
                                               0.97
                          0.97
                                     0.97
                                                         2445
[]: from sklearn.svm import SVC
[]: svc_params = {'kernel': ['rbf'], 'gamma': [0.01, 0.001, 0.0001], 'C': [1, 10, ___
     →100]}
     clf = GridSearchCV(estimator=SVC(), param_grid=svc_params, cv=3, n_jobs=-1,
                        scoring='accuracy', verbose=10)
[]: clf.fit(X_for_SVM, y_train_labels_encoded)
    Fitting 3 folds for each of 9 candidates, totalling 27 fits
[]: GridSearchCV(cv=3, estimator=SVC(), n_jobs=-1,
                  param_grid={'C': [1, 10, 100], 'gamma': [0.01, 0.001, 0.0001],
                              'kernel': ['rbf']},
                  scoring='accuracy', verbose=10)
[]: prediction_svm1 = clf.predict(X_test_features)
     #Inverse le transform to get original label back.
     prediction_svm1 = le.inverse_transform(prediction_svm1)
[]: print ("Accuracy = ", metrics.accuracy_score(y_test, prediction_svm1)*100)
     cm = confusion_matrix(y_test, prediction_svm1)
     print(cm)
     cm_df = pd.DataFrame(cm, index=le.classes_, columns=le.classes_)
     cm_df.head()
    Accuracy = 97.70961145194273
    ΓΓ1076
                   31
             15
     [ 20
            906
                   7]
             11 407]]
     0
[]:
                      Covid Normal Viral_Pneumonia
     Covid
                       1076
                                 15
                                                   3
                                                   7
                                906
     Normal
                         20
     Viral_Pneumonia
                                                 407
                          0
                                 11
[]: plt.figure(figsize=(9,9))
     sns.set(font_scale=1.5, color_codes=True, palette='deep')
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```
sns.heatmap(cm_df, annot=True, annot_kws={'size':16}, fmt='d', cmap='Pastel1')
plt.ylabel("True Label")
plt.xlabel("Predicted Label")
plt.title('Confusion Matrix of Validation Data', size=15)
plt.show()
```



[]: print(classification_report(y_test, prediction_svm1))

support	f1-score	recall	precision	
1094	0.98	0.98	0.98	Covid
933	0.97	0.97	0.97	Normal

```
0.97
    Viral_Pneumonia
                           0.98
                                               0.97
                                                           418
                                               0.98
                                                         2445
           accuracy
          macro avg
                           0.98
                                     0.98
                                               0.98
                                                         2445
       weighted avg
                           0.98
                                     0.98
                                               0.98
                                                         2445
[]: | img5= cv2.imread(r'C:\Users\josep\Desktop\covid\Oovid19-dataset\test\Covid\098.
     →png', cv2.IMREAD_COLOR)
     img5 = cv2.cvtColor(img5, cv2.COLOR_BGR2RGB)
     img5 = cv2.resize(img5, (224,224))/255.0
     input_img = np.expand_dims(img5, axis=0) #Expand dims so the input is (num_
      \rightarrow images, x, y, c)
     img_test = VGG_model.predict(input_img)
     input_img_features=img_test.reshape(img_test.shape[0], -1)
     pred = clf.predict(input_img_features)
     pred = le.inverse_transform([pred])
    pred
    C:\Users\josep\AppData\Local\Programs\Python\Python38\lib\site-
    packages\sklearn\utils\validation.py:63: DataConversionWarning: A column-vector
    y was passed when a 1d array was expected. Please change the shape of y to
    (n_samples, ), for example using ravel().
      return f(*args, **kwargs)
[]: array(['Covid'], dtype='<U15')
```

[]: