```
from google.colab import drive
drive.mount('/content/drive')
```

```
Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount
```

```
from numpy import mean
from numpy import std
from matplotlib import pyplot
from sklearn.model_selection import KFold
from keras.datasets import mnist
from keras.utils import to_categorical
from keras.models import Sequential
from keras.layers import Conv2D
from keras.layers import MaxPooling2D
from keras.layers import Dense
from keras.layers import Flatten
from keras.optimizers import SGD
from keras.layers import Dropout
from keras.layers import BatchNormalization
import keras
from keras import backend as K
import matplotlib.pyplot as plt
import sklearn


path_normal = '/content/drive/MyDrive/Deep learning demo project/Normal/'
path_pneumonia = '/content/drive/MyDrive/Deep learning demo project/pneumonia/'


##Import necessary libraries
import numpy as np
import PIL
import cv2
import os
data1 = list()
data2 = list()
x = list()


##Class-1 images##

for image in os.walk(path_normal):
  data1.append(image[2])

for i in range(len(data1[0])):
  str_complete = path_normal + data1[0][i]
  img = cv2.imread(str_complete)
  img = cv2.resize(img, (224, 224))
  x.append(img)
  print(i)#Ensure all images are loaded
```

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```
print(img.shape)
```

```
(224, 224, 3)
```

```
##Class-2 images##

for image in os.walk(path_pneumonia):
  data2.append(image[2])

for i in range(len(data2[0])):
  str_complete = path_pneumonia + data2[0][i]
  img = cv2.imread(str_complete)
  img = cv2.resize(img, (224, 224))
  x.append(img)#Ensure all images are loaded
  print(i)
```

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```

```
data_x = np.asarray(x)
```

```
data_x.shape
```

```
(1400, 224, 224, 3)
```

```
x=data_x
```

```
y = np.zeros(1400)
y[:700] =1
y[700:1400]=2
```

```
from sklearn.model_selection import train_test_split
```

```
##Dataset Split##
from sklearn.model_selection import train_test_split
from keras.utils import to_categorical
#y = to_categorical(y)
x_train, x_test, y_train, y_test = train_test_split(data_x, y, test_size=0.2, random_state=1)
x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_size=1/8, random_state=1)

y_tr_one_hot  = np.zeros((np.array(y_train).shape[0],2))

for i in range(np.array(y_train).shape[0]):
  label = y_train[i]-1
  y_tr_one_hot[i][int(label)] = 1
```

```
y_val_one_hot  = np.zeros((np.array(y_val).shape[0],2))

for i in range(np.array(y_val).shape[0]):
  label = y_val[i]-1
  y_val_one_hot[i][int(label)] = 1
```

```
y_te_one_hot  = np.zeros((np.array(y_test).shape[0],2))

for i in range(np.array(y_test).shape[0]):
  label = y_test[i]-1
  y_te_one_hot[i][int(label)] = 1
```

```
from keras.models import load_model
from keras.layers import Lambda
import tensorflow as tf
from tensorflow.keras.models import Model
from tensorflow.keras.layers import GlobalAveragePooling2D, Dense
```

```
model = tf.keras.applications.ResNet152V2(include_top=False,input_shape=(224,224,3))
# mark loaded layers as not trainable
for layer in model.layers:
  layer.trainable = False
# add new classifier layers
flat1 = Flatten()(model.layers[-1].output)
#x=Dense(1024,activation='relu')(flat1) # FC layer 1
#x=Dense(64,activation='relu')(x)   # FC layer 2
output = Dense(2, activation='softmax')(flat1)
model = Model(inputs=model.inputs, outputs=output)

optimizer = tf.keras.optimizers.Adam(learning_rate=0.01)
model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
model.fit(x_train, y_tr_one_hot, validation_data=(x_val, y_val_one_hot), epochs=5, batch_size=200,verbose=1)
```

```
Epoch 1/5
5/5 [==============================] - 610s 123s/step - loss: 5734.9819 - accuracy: 0.5102 - val_loss: 3202.9597 - val_ac
```

```
Epoch 2/5
5/5 [==============================] - 598s 123s/step - loss: 1401.5342 - accuracy: 0.6786 - val_loss: 1154.0453 - val_ac
Epoch 3/5
5/5 [==============================] - 598s 123s/step - loss: 345.1577 - accuracy: 0.8622 - val_loss: 561.3847 - val_accu
Epoch 4/5
5/5 [==============================] - 606s 122s/step - loss: 312.5118 - accuracy: 0.8847 - val_loss: 53.5582 - val_accur
Epoch 5/5
5/5 [==============================] - 617s 126s/step - loss: 99.0872 - accuracy: 0.9653 - val_loss: 207.2567 - val_accur
<keras.callbacks.History at 0x7f07398b7f10>
```

```
import sklearn
from sklearn.metrics import confusion_matrix
```

```
test_loss, test_acc = model.evaluate(np.array(x_test), np.array(y_te_one_hot), verbose=0)
print(test_acc)
##Evaluating Sensitivity, Accuracy and Kappa scores
y_prob = model.predict(x_test)
Y_pred = y_prob.argmax(axis=-1)
```

```
0.9607142806053162
9/9 [==============================] - 141s 15s/step
```

```
cm1 = confusion_matrix(y_test-1,Y_pred)
print("confusion matrix \n",cm1)
```

```
confusion matrix
 [[126  11]
 [  0 143]]
```

```
from sklearn.metrics import classification_report
```

```
import pandas as pd
```

```
print(pd.DataFrame(classification_report(y_test-1,Y_pred,output_dict=True)).T)
Kappa=sklearn.metrics.cohen_kappa_score(y_test-1,Y_pred)
print('Kappa=',Kappa)
```

```
                precision    recall  f1-score     support
0.0              1.000000  0.919708  0.958175  137.000000
1.0              0.928571  1.000000  0.962963  143.000000
accuracy         0.960714  0.960714  0.960714    0.960714
macro avg        0.964286  0.959854  0.960569  280.000000
weighted avg     0.963520  0.960714  0.960620  280.000000
Kappa= 0.9212598425196851
```

✓ 0s completed at 09:51