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KIDNEY STONE DETECTION USING DIP

Submitted as the **Content Beyond the Syllabus** for the subject Computer Graphics and Visualization (18CS62)

Submitted by

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1. INTRODUCTION:

A kidney stone, also known as renal calculi, is a hard mass or crystal-like deposit that forms in the kidneys. It is composed of various substances, such as calcium, oxalate, uric acid, or cystine, that are present in urine. Kidney stones can vary in size, ranging from as small as a grain of sand to as large as a golf ball. The exact cause of kidney stone formation is not always clear, but several factors can contribute to their development, including:

- Dehydration: Insufficient fluid intake can lead to concentrated urine, increasing the risk of stone formation.
- Diet: Consuming foods high in oxalate, such as spinach, rhubarb, chocolate, and certain nuts, can contribute to calcium oxalate stones. A high-sodium or high-protein diet may also increase the risk
- Family history: A person with a family history of kidney stones is more likely to develop them
- Certain medical conditions: Conditions like hyperparathyroidism, gout, urinary tract infections, and certain digestive disorders can increase the risk of kidney stones.

1.1 KIDNEY STONE DETECTION USING DIP

Here are some general steps that can be considered:

- 1. Image Acquisition: Obtain images of the kidneys or urinary tract using an appropriate imaging modality such as ultrasound or CT scan. These images will serve as input for the subsequent DIP analysis.
- 2. Preprocessing: Apply preprocessing techniques to enhance the quality of the acquired images. This may involve noise reduction, contrast enhancement, and image filtering to improve the clarity of the kidney stone regions.
- 3. Segmentation: Segment the kidney stones from the background and surrounding tissues in the acquired images. Various segmentation algorithms can be employed, such as thresholding, edge detection, region-growing, or clustering methods, to separate the stone regions of interest.
- 4. Feature Extraction: Extract relevant features from the segmented kidney stone regions. These features could include shape descriptors, texture features, or statistical properties that capture the characteristics of the stones.
- 5. Classification: Utilize a classification algorithm or machine learning technique to classify the extracted features as kidney stones or non-stone regions. This could involve training a classifier using a labeled dataset to distinguish between normal kidney tissue and different types of stones.

2.Concept/Algorithm:

2.1 VGG-16 Model

- VGG16 is a type of CNN (Convolutional Neural Network) that is considered to be one of the best computer vision models to date.
- VGG16 is object detection and classification algorithm which is able to classify 1000 images of 1000 different categories with 92.7% accuracy. It is one of the popular algorithms for image classification and is easy to use with transfer learning.
- The 16 in VGG16 refers to 16 layers that have weights. In VGG16 there
 are thirteen convolutional layers, five Max Pooling layers, and three Dense layers
 which sum up to 21 layers but it has only sixteen weight layers i.e., learnable
 parameters layer.
- VGG16 takes input tensor size as 224, 244 with 3 RGB channel
- Most unique thing about VGG16 is that instead of having a large number of hyperparameters they focused on having convolution layers of 3x3 filter with stride 1 and always used the same padding and maxpool layer of 2x2 filter of stride 2
- The convolution and max pool layers are consistently arranged throughout the whole architecture
- Conv-1 Layer has 64 number of filters, Conv-2 has 128 filters, Conv-3 has 256 filters, Conv 4 and Conv 5 has 512 filters.
- Three Fully-Connected (FC) layers follow a stack of convolutional layers: the first two have 4096 channels each, the third performs 1000-way ILSVRC classification and thus contains 1000 channels (one for each class). The final layer is the soft-max layer.

3. CODE

```
# Import libraries, packages, modules, functions, etc...
import numpy as np
import cv2
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Activation, Dense, Flatten,
BatchNormalization, Conv2D, MaxPool2D, Dropout
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.metrics import categorical crossentropy
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from sklearn.metrics import confusion matrix
import itertools
import os
import shutil
import random
import glob
import matplotlib.pyplot as plt
from sklearn.preprocessing import MinMaxScaler
import warnings
from sklearn.utils import shuffle
from sklearn.model_selection import train_test split
import skimage
from skimage.restoration import denoise nl means, estimate sigma,
denoise tv_chambolle, denoise_bilateral,denoise_wavelet, unsupervised_wiener
from scipy.signal import convolve2d as conv2
from skimage.filters import median
import copy
from tqdm import tqdm
from scipy import ndimage as nd
import multiprocessing
from skimage.metrics import peak_signal_noise_ratio
from skimage.metrics import mean squared error
from skimage.metrics import structural similarity as ssim
from itertools import repeat
import zipfile
import os
from IPython.display import FileLink
from sklearn.svm import SVC
from keras.callbacks import EarlyStopping
import pandas as pd
from sklearn.cluster import KMeans
from sklearn import metrics
from sklearn.cluster import MiniBatchKMeans
from sklearn.metrics import accuracy score
from tensorflow.keras.preprocessing import image
from skimage.morphology import disk
from tensorflow.keras.optimizers import SGD
warnings.simplefilter(action='ignore', category=FutureWarning)
!pip install natsort
!pip install MedPy
import natsort
from medpy.filter.smoothing import anisotropic diffusion
```

%matplotlib inline

```
Requirement already satisfied: natsort in c:\users\prajwal
mr\anaconda3\lib\site-packages (8.4.0)
Requirement already satisfied: MedPy in c:\users\prajwal mr\anaconda3\lib\site-
packages (0.4.0)
Requirement already satisfied: numpy>=1.11.0 in c:\users\prajwal
mr\anaconda3\lib\site-packages (from MedPy) (1.22.4)
Requirement already satisfied: scipy>=1.1.0 in c:\users\prajwal
mr\anaconda3\lib\site-packages (from MedPy) (1.7.3)
Requirement already satisfied: SimpleITK>=1.1.0 in c:\users\prajwal
mr\anaconda3\lib\site-packages (from MedPy) (2.2.1)
                                                                            In [2]:
#pip install opencv-python
                                                                            In [3]:
#pip install tensorflow
                                                                            In [4]:
# Get the complete data of kidney stone images
import os
def Get data(dir, catagories, data):
    for category in catagories:
        path = os.path.join(dir, category)
        class_number = catagories.index(category)
        for img in os.listdir(path):
            trv:
                img_array = cv2.imread(os.path.join(path , img))
                new image = cv2.resize(img array,(128,128))
                data.append([new_image , class_number])
            except Exception as e:
                pass
data = []
# Augmented Dataset
# Get_data("/kaggle/input/combined-aug-ks/KS_Detection", ["Combined_N",
"Combined KS"], data)
# Original Dataset
Get_data("C:/Users/prajwal mr/Downloads/archive/CT_SCAN", ["Kidney_stone",
"Normal"], data)
                                                                            In [5]:
catagories = ["Kidney_stone", "Normal"]
print(catagories.index("Kidney stone"))
print(catagories.index("Normal"))
0
1
                                                                            In [6]:
print('Number of kidney stone images',len(os.listdir('C:/Users/prajwal
mr/Downloads/archive/CT SCAN/Kidney stone')))
print('Number of Normal kidney images',len(os.listdir('C:/Users/prajwal
mr/Downloads/archive/CT_SCAN/Normal')))
```

```
Number of Normal kidney images 828
                                                                                    In [7]:
print(data[143][0].shape)
plt.imshow(data[143][0])
plt.show()
(128, 128, 3)
  20
  40
  60
  80
 100
 120
         20
              40
                   60
                        80
                            100
                                 120
    0
                                                                                    In [8]:
# Extract Features and Labels from the data
Features = []
Labels = []
for features , labels in data:
    Features.append(features)
    Labels.append(labels)
Features = np.array(Features)
Labels = np.array(Labels)
                                                                                    In [9]:
Features[5][1]
                                                                                   Out[9]:
array([[
           0,
                 0,
                      0],
               88,
                     88],
          88,
                 0,
                      0],
           0,
                 0,
           0,
                      0],
           2,
                 2,
                      2],
           6,
                 6,
                      6],
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                 0,
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                      0],
           0,
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                 1,
                      1],
           4,
                4,
                      4],
          15,
               15,
                     15],
          90,
               90,
                     90],
           1,
                 1,
                      1],
                 0,
                      0],
```

Number of kidney stone images 781

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             75],
[113, 113, 113],
[118, 118, 118],
[121, 121, 121],
[106, 106, 106],
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[ 59,
       59,
[ 77, 77, 77],
[100, 100, 100],
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               0],
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         0,
               0],
               0]], dtype=uint8)
         0,
   0,
```

In [10]:

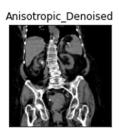
```
print(Labels.shape)
(1609, 128, 128, 3)
(1609,)
                                PREPROCESSING
                                                                           In [11]:
def Processed(Processed images):
    img per row = 4
    fig, ax = plt.subplots(nrows = 2, ncols = img per row, figsize=(10,10),
subplot_kw = dict(xticks = [], yticks = []))
    for row in [0, 1]:
        for col in range(img_per_row):
            if(row * img_per_row + col == 7):
                break
            if(Processed_images[row * img_per_row + col][1] ==
"Bilateral_Denoised" or Processed_images[row * img_per_row + col][1] ==
"Wavelet_Denoised" or Processed_images[row * img_per_row + col][1] ==
"Non Local Means"):
                ax[row, col].imshow(Processed_images[row * img_per_row +
col][0].astype('float64'), cmap = "gray")
            else:
                ax[row, col].imshow(Processed_images[row * img_per_row +
col][0].astype('uint8'), cmap = "gray")
            ax[row, col].set_title(Processed_images[row * img_per_row +
col][1])
            plt.axis('off')
    plt.show()
                                                                           In [12]:
# Filters
# Efficient in reducing gaussian image
# Lowers the image resolution
def AnisotropicFilter_wholedataset(noised_dataset, niter, kappa, gamma, option,
size):
    anisotropic dataset = copy.deepcopy(noised dataset)
    for i in tqdm(range(anisotropic dataset.shape[0])):
        anisotropic_dataset[i] = anisotropic_diffusion(noised_dataset[i, :, :,
:], niter=niter, kappa=kappa, gamma=gamma, option=option).reshape(size, size,
1)
    return anisotropic_dataset
# Reduce spiky noise
# Non linear filter
# Smoothens the image
# Prevent blur and preserve sharp edges
# Only better for removing salt and pepper noise
def median wholedataset(noised dataset, filtersize, size):
    median wholedata = copy.deepcopy(noised dataset)
    for i in tqdm(range(noised dataset.shape[0])):
        median_wholedata[i] = median(noised_dataset[i,:,:,:][:,:,0],
np.ones((filtersize, filtersize))).reshape(size, size, 1)
    return median wholedata
```

```
# Regardless of the frequency composition of the signal, eliminate noise while
keeping its properties
# Unable to preserve fine details in case of high noise data.
# There's a chance the wavelet coefficients are biased
def wavelet_wholedataset(noised_dataset, sigma, wavelet_levels, size):
    wavelet wholedata = copy.deepcopy(noised dataset)
    for i in tqdm(range(noised_dataset.shape[0])):
        wavelet_wholedata[i] = denoise_wavelet(noised_dataset[i,:,:,:],
sigma=sigma, channel axis=-1, wavelet levels=wavelet levels,
rescale_sigma=True).reshape(size, size, 1)
    return wavelet wholedata
# Better in preserving edges
# Effectively remove gaussian noise
# Not the best fit for salt and pepper noise
def BilateralFilter wholedataset(noised dataset, sigma color, sigma spatial,
channel_axis, size):
    bilateral_dataset = copy.deepcopy(noised_dataset)
    for i in tqdm(range(bilateral dataset.shape[0])):
        bilateral dataset[i] = denoise bilateral(noised dataset[i, :, :, :],
sigma_color = sigma_color, sigma_spatial=sigma_spatial,
channel_axis=channel_axis).reshape(size, size, 1)
    return bilateral dataset
# Effect for gaussian noise
# Simplicity of algorithm is one of the pros
# Reduce the image details
# Unable to preserve edges
# Denoised images would be blurry.
def GaussianFilter wholedataset(noised dataset, sigma):
    gaussian_dataset = copy.deepcopy(noised_dataset)
    for i in tqdm(range(gaussian_dataset.shape[0])):
        gaussian dataset[i] = nd.gaussian filter(tuple(noised dataset[i, :, :,
:]), sigma=sigma)
    return gaussian_dataset
# Preserve edges
# Better performance with redundant images
# Expensive, so non-suggestable for larger noise
def non local mean(image):
    sigma_est = np.mean(estimate_sigma(image, channel_axis=-1))
    patch_kw = dict(patch_size = 5,  # 5x5 patches
                patch_distance = 6, # 13x13 search area
                channel axis = -1)
    denoise fast = denoise nl means(image, h = 0.6 * sigma est,
sigma=sigma_est, fast_mode=True, **patch_kw)
    return denoise fast
# Apply bilateral filter with d = 15
# sigmaColor = sigmaSpace = 75
# d: Diameter of each pixel neighborhood.
# sigmaColor: Value of sigma in the color space. The greater the value, the
```

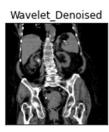
```
colors farther to each other will start to get mixed.
# sigmaSpace: Value of sigma in the coordinate space. The greater its value,
the more further pixels will mix together, given that their colors lie within
the sigmaColor range.
def cv2bilateralFilter(image, d, sigmaColor, sigmaSpace):
    bilateral = cv2.bilateralFilter(image, d, sigmaColor, sigmaSpace)
    bilateral = bilateral.astype('uint8')
    return bilateral
# Execution
# ani = AnisotropicFilter wholedataset(noised dataset, 50, 20, 0.2, 1)
# bi = BilateralFilter_wholedataset(noised_dataset, 15, -1)
# gauss = GaussianFilter_wholedataset(noised_dataset, 2)
# median_img = median(Features[0, :, :, :][:,:,0], disk(3), mode = 'constant',
cval = 0.0).reshape(128, 128, 1)
# wavelet_img = denoise_wavelet(Features[0, :, :, :], sigma = 0.12,
channel axis = -1,convert2ycbcr=True, method='BayesShrink', mode='soft',
rescale_sigma=True).reshape(128, 128, 3)
# bilateral_image = denoise_bilateral(Features[0], sigma_spatial = 10,
channel axis = -1)
# anisotropic_image = anisotropic_diffusion(Features[700,:,:,:][:,:,0], niter =
100, kappa = 10, gamma = 0.02, option=1)
# anisotropic image = anisotropic diffusion(Features[700,:,:,:][:,:,0], niter =
50, kappa = 10, gamma = 0.02, option=1)
# anisotropic_image = anisotropic_diffusion(Features[60,:,:,:][:,:,0], niter =
50, kappa = 5, gamma = 0.005, option=1)
# anisotropic_image = anisotropic_diffusion(Features[600,:,:,:][:,:,0], niter =
50, kappa = 5, gamma = 0.000001, option=1)
# anisotropic image = anisotropic image.astype('uint8')
# Metrics
def psnr wholedataset(dataset original, dataset denoised):
    sumpsnr = 0
    avgpsnr = 0
    for i in tqdm(range(dataset original.shape[0])):
        true_min, true_max = np.min(dataset_original[i, :, :, :]),
np.max(dataset original[i, :, :, :])
        dataRange = abs(true min)+abs(true max)
        psnr = peak_signal_noise_ratio(dataset_original[i],
dataset_denoised[i], data_range=dataRange)
        sumpsnr += psnr
    avgpsnr = sumpsnr/dataset_original.shape[0]
    return avgpsnr
# psnr_wholedataset(data, noised_dataset)
# Execution
# true min, true max = np.min(Features[0]), np.max(Features[0])
# dataRange = abs(true_min) + abs(true_max)
```

```
# print("PSNR: ", peak_signal_noise_ratio(Features[143, :, :, :][:,:,0],
anisotropic image, data range = dataRange))
# print("MSE: ", mean squared error(Features[143,:,:,:][:,:,0],
anisotropic image))
# print("SSIM: ", ssim(Features[143, :, :, :][:,:,0], anisotropic image,
multichannel=True, gaussian weights = True, sigma=1.5,
use sample covariance=False))
# plt.imshow(np.hstack((wavelet_img, Features[143, :, :,
:].reshape(128,128,3))), cmap = "gray")
                                                                          In [13]:
true min, true max = np.min(Features[143]), np.max(Features[143])
dataRange = abs(true min) + abs(true max)
bilateral = cv2bilateralFilter(Features[143,:,:,:], 15, 75, 75)
print("PSNR: ", peak signal noise ratio(Features[143,:,:,:], bilateral,
data range = dataRange))
print("MSE: ", mean_squared_error(Features[143,:,:,:], bilateral))
print("SSIM: ", ssim(Features[143,:,:,:], bilateral, multichannel = True,
gaussian_weights = True, sigma=1.5, use_sample_covariance=False))
PSNR: 31.786955991118436
MSE: 43.0908203125
SSIM: 0.9190078489547154
                                                                          In [14]:
anisotropic image = anisotropic diffusion(Features[143,:,:,:][:,:,0], niter =
50, kappa = 5, gamma = 0.005, option=1)
bilateral image = denoise bilateral(Features[143], sigma spatial = 10,
channel axis = -1)
wavelet img = denoise wavelet(Features[143, :, :, :], sigma = 0.12,
channel axis = -1,convert2ycbcr = True, method = 'BayesShrink', mode = 'soft',
rescale sigma = True).reshape(128, 128, 3)
median_img = median(Features[143, :, :, :][:,:,0], disk(3), mode = 'constant',
cval = 0.0).reshape(128, 128, 1)
guassian img = nd.gaussian filter(tuple(Features[143, :, :, :]), sigma = 2)
non local mean img = non local mean(Features[143])
Processed_images = [[Features[143, :, :, :], "Original_Image"],
[anisotropic image, "Anisotropic Denoised"], [bilateral image,
"Bilateral Denoised"], [wavelet_img, "Wavelet_Denoised"],
[median_img, "Median_Denoised"], [guassian_img, "Gaussian_Denoised"],
[non_local_mean_img, "Non_Local_Means"]]
Processed(Processed images)
```







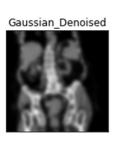




processed.

def edge detection(Image):

img = cv2.Canny(Image, 100, 200)





```
In [15]:
# Denoising the given data with anisotropic Filter
true min, true max = np.min(Features[0]), np.max(Features[0])
dataRange = abs(true min) + abs(true max)
anis_denoised = []
for i in range(len(Features)):
    anisotropic image = anisotropic diffusion(Features[i,:,:,:], niter = 100,
kappa = 10, gamma = 0.02, option = 1)
    anis_denoised.append(anisotropic_image)
anis_denoised = np.array(anis_denoised)
          #Cropping, Thersholding(or masking), Edge Detection, Morphological Analysis
                                                                            In [16]:
def crop(image_to_be_cropped):
    height, width = image to be cropped.shape[:2]
    start row, start col = int(height * .24), int(width * .24)
    end_row, end_col = int(height * .78), int(width * .78)
#
      start_row, start_col = int(height * .24), int(width * .24)
      end_row, end_col = int(height * .42), int(width * .78)
    cropped image = image to be cropped[start row:end row, start col:end col]
    return cropped image
def threshold(img, thresh1 = 254):
    return ((img > thresh1) * 255).astype('uint8')
# Canny edge detection is a technique to extract useful structural information
from different vision objects and dramatically reduce the amount of data to be
```

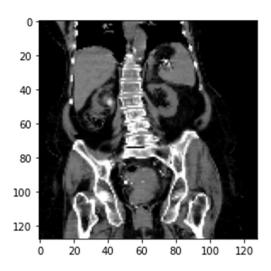
```
return img
```

```
# To see the pixel values division in the images
def hist_plot(Image):
    plt.hist(Image.flat, bins = 100, range = (0, 255))
    plt.show()

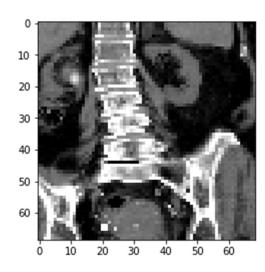
# Convert BGR image to Gray Scale image
gray_image = cv2.cvtColor(data[143][0], cv2.COLOR_BGR2GRAY)

# expanding dimensions from 128,128 to 128,128,1
gray_image = np.expand_dims(gray_image, axis = -1)
```

plt.imshow(gray_image, cmap = "gray")
plt.show()



Crop a Image
cropped_image = crop(Features[143,:,:,:])
plt.imshow(cropped_image)
plt.show()



In [19]:

In [18]:

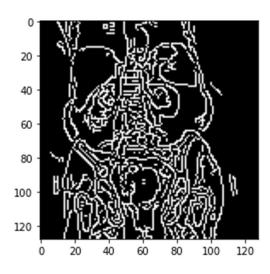
In [17]:

Out[19]:

In [20]:

(69, 69, 3)

Edge Detection img = edge detection(gray image) plt.imshow(img, cmap = "gray") plt.show()



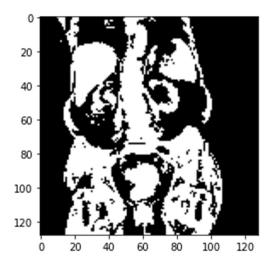
In [21]:

Threshold masking using OTSU method

Automatic Thresholding

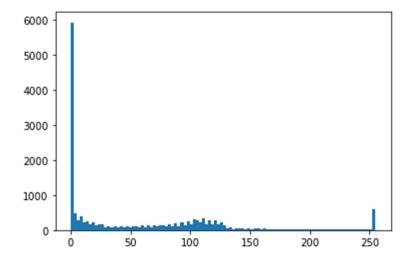
Otsu's method looks at every possible value for the threshold between background and foreground, calculates the variance within each of the two clusters, and selects the value for which the weighted sum of these variances is the least.

re, th = cv2.threshold(gray_image, 0, 255, cv2.THRESH_BINARY + cv2.THRESH_OTSU) plt.imshow(th, cmap = "gray") plt.show()



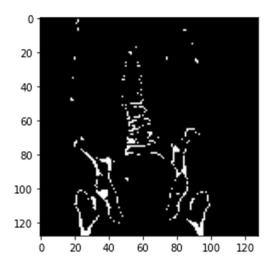
In [22]:

See the pixel values division in the images hist_plot(gray_image)



In [23]:

Manual Thresholding on the basis of pixel divisions
img = threshold(gray_image)
plt.imshow(img, cmap = "gray")
plt.show()



In [24]:

Mask of Image

HSV or Hue Saturation Value is used to separate image luminance from color information. This makes it easier when we are working on or need luminance of the image/frame. HSV also used in situations where color description plays an integral role.

Hue, in the context of color and graphics, refers to the attribute of a visible light due to which it is differentiated from or similar to the primary colors

Saturation describes the intensity of the color.

Hue is determined by the dominant wavelength of the visible spectrum. It is the attribute that permits colors to be classified as red, yellow, green, blue, or an intermediate color. Saturation pertains the amount of white light mixed with a hue.

Value works in conjunction with saturation and describes the brightness or intensity of the color, from 0 to 100 percent, where 0 is completely black, and 100 is the brightest and reveals the most color.

lower_yellow = np.array([35, 255, 255])

```
upper_yellow = np.array([25, 50, 70])
lower_blue = np.array([110,50,50])
upper_blue = np.array([130,255,255])
lower_white = np.array([180, 18, 255])
upper_white = np.array([0, 0, 231])
lower_black = np.array([180, 255, 30])
upper black = np.array([0, 0, 0])
bgr_image = cv2.cvtColor(Features[143,:,:][:,:,0], cv2.COLOR_GRAY2BGR)
hsv = cv2.cvtColor(bgr_image, cv2.COLOR_BGR2HSV)
mask = cv2.inRange(hsv, lower black, upper black)
detected output = cv2.bitwise and(hsv, hsv, mask = mask)
plt.imshow(detected_output, cmap = "gray")
plt.show()
  0
  20
  40
  60
  80
 100
 120
        20
            40
                 60
                     80
                          100
                              120
                                                                            In [25]:
# Crop all the input images
Cropped_Features = []
for i in range(len(Features)):
    cropped image = crop(Features[i,:,:,:])
    Cropped Features.append(cropped image)
Cropped_Features = np.array(Cropped_Features)
                                                                            In [26]:
# Denoise all the input images with Bilateral Filte
# On original images
bil_de = []
for i in range(len(Features)):
    bilateral_de = cv2.bilateralFilter(Features[i], 15, 75, 75)
    bil de.append(bilateral de)
bil_de = np.array(bil_de)
# On cropped images
bilateral_denoised = []
for i in range(len(Cropped_Features)):
    bilateral = cv2.bilateralFilter(Cropped Features[i], 15, 75, 75)
    bilateral denoised.append(bilateral)
```

```
bilateral_denoised = np.array(bilateral_denoised)
                                                                            In [27]:
# Threshold masking on all the input images
# Original
thresholded = []
for i in range(len(bilateral denoised)):
    temp = threshold(bilateral_denoised[i])
    thresholded.append(temp)
thresholded = np.array(thresholded)
# Cropped
thresholded = []
for i in range(len(Cropped_Features)):
    temp = threshold(Cropped Features[i])
    thresholded.append(temp)
thresholded = np.array(thresholded)
      SPLITTING DATA
                                                                            In [28]:
print(Features.shape)
print(Labels.shape)
(1609, 128, 128, 3)
(1609,)
                                                                            In [29]:
Xtrain, Xtest, Ytrain, Ytest = train_test_split(Features, Labels, test_size =
0.30, random_state = 80, shuffle = True)
print(len(Xtrain), len(Xtest), len(Ytrain), len(Ytest))
# Converting the list to a numpy array as a requirement for the input in fit
function.
Xtrain=np.array(Xtrain)
Xtest=np.array(Xtest)
Ytrain=np.array(Ytrain)
Ytest=np.array(Ytest)
1126 483 1126 483
                                                                             In [ ]:
                                                                            In [49]:
from tensorflow.keras.optimizers import SGD
                                                                            In [50]:
model3 = Sequential()
model3.add(Conv2D(32, (3,3), activation='relu', input shape=(128, 128, 3),
padding='same'))
model3.add(MaxPool2D(2))
model3.add(Dropout(0.2))
```

```
model3.add(Conv2D(32, (3,3), activation='relu', padding='same'))
model3.add(MaxPool2D(pool_size=(2,2), strides = 2))
model3.add(Dropout(0.4))
model3.add(Conv2D(64, (3,3), activation='relu', padding='same'))
model3.add(MaxPool2D(pool_size=(2,2), strides = 2))
model3.add(Dropout(0.5))
model3.add(Conv2D(64, (3,3), activation='relu', padding='same'))
model3.add(MaxPool2D(pool_size=(2,2), strides = 2))
model3.add(Conv2D(128, (3,3), activation='relu', padding='same'))
model3.add(MaxPool2D(pool size=(2,2), strides = 2))
model3.add(Conv2D(128, (3,3), activation='relu', padding='same'))
model3.add(MaxPool2D(pool_size=(2,2), strides = 2))
model3.add(Dropout(0.6))
model3.add(Flatten())
model3.add(Dense(512, activation='relu'))
model3.add(Dense(1, activation='sigmoid'))
model3.compile(loss = "binary_crossentropy", optimizer =
Adam(learning_rate=0.001), metrics=['accuracy'])
      Different types of Model Training and Testing
CONVOLUTIONAL NEURAL NETWORK(CNN)
                                                                           In [ ]:
                                                                          In [51]:
# verbose=2 just specifies how much output to the console we want to see during
each epoch of training. The verbosity levels range from 0 to 2, so we're
getting the most verbose output.
# model.fit(Xtrain, Ytrain, epochs = 50, batch size = 20, verbose = 2)
# Define a callback for early stopping
early stopping = EarlyStopping(monitor = 'val loss', patience = 10)
# Train the model
history = model3.fit(Xtrain, Ytrain, epochs = 200, batch_size = 20, callbacks =
[early_stopping])
#To this function, we pass in the test samples x, specify a batch size, and
specify which level of verbosity we want from log messages during prediction
generation. The output from the predictions won't be relevant for us, so we're
setting verbose=0 for no output.
#Note that, unlike with training and validation sets, we do not pass the labels
of the test set to the model during the inference stage.
predictions = model3.predict(x = Xtest, batch_size = 20, verbose = 0)
Accuracy = model3.evaluate(Xtest, Ytest, verbose = 0)
print("Accuracy: ", Accuracy[1] * 100)
Epoch 1/200
57/57 [============= ] - ETA: 0s - loss: 9.2786 - accuracy:
0.4787WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
```

```
is not available. Available metrics are: loss,accuracy
accuracy: 0.4787
Epoch 2/200
0.5080WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5080
Epoch 3/200
57/57 [========================== ] - ETA: 0s - loss: 0.6944 - accuracy:
0.5169WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5169
Epoch 4/200
0.5266WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5266
Epoch 5/200
57/57 [============== ] - ETA: 0s - loss: 0.6980 - accuracy:
0.5160WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss, accuracy
57/57 [============ ] - 19s 332ms/step - loss: 0.6980 -
accuracy: 0.5160
Epoch 6/200
0.5027WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5027
Epoch 7/200
57/57 [============ ] - ETA: 0s - loss: 0.6969 - accuracy:
0.5098WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5098
Epoch 8/200
0.5169WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5169
Epoch 9/200
0.5053WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5053
Epoch 10/200
57/57 [==================== ] - ETA: 0s - loss: 0.6921 - accuracy:
0.5249WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5249
```

```
Epoch 11/200
0.5169WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5169
Epoch 12/200
57/57 [============ ] - ETA: 0s - loss: 0.6928 - accuracy:
0.5187WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5187
Epoch 13/200
0.5355WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5355
Epoch 14/200
57/57 [========================= ] - ETA: 0s - loss: 0.6939 - accuracy:
0.5169WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5169
Epoch 15/200
0.5231WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 20s 346ms/step - loss: 0.6921 -
accuracy: 0.5231
Epoch 16/200
57/57 [========================== ] - ETA: 0s - loss: 0.6958 - accuracy:
0.5151WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 20s 352ms/step - loss: 0.6958 -
accuracy: 0.5151
Epoch 17/200
57/57 [======================== ] - ETA: 0s - loss: 0.6931 - accuracy:
0.5018WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5018
Epoch 18/200
57/57 [========================= ] - ETA: 0s - loss: 0.6883 - accuracy:
0.5471WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 20s 353ms/step - loss: 0.6883 -
accuracy: 0.5471
Epoch 19/200
57/57 [========================= ] - ETA: 0s - loss: 0.6944 - accuracy:
0.5115WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5115
Epoch 20/200
57/57 [========================= ] - ETA: 0s - loss: 0.6940 - accuracy:
0.5329WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
```

```
is not available. Available metrics are: loss,accuracy
57/57 [================= ] - 30s 538ms/step - loss: 0.6940 -
accuracy: 0.5329
Epoch 21/200
0.5382WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5382
Epoch 22/200
57/57 [========================== ] - ETA: 0s - loss: 0.6925 - accuracy:
0.5240WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============================ ] - 41s 712ms/step - loss: 0.6925 -
accuracy: 0.5240
Epoch 23/200
0.5151WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5151
Epoch 24/200
57/57 [============== ] - ETA: 0s - loss: 0.6897 - accuracy:
0.5391WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss, accuracy
57/57 [============ ] - 38s 670ms/step - loss: 0.6897 -
accuracy: 0.5391
Epoch 25/200
0.5320WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5320
Epoch 26/200
57/57 [============ ] - ETA: 0s - loss: 0.6934 - accuracy:
0.5142WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5142
Epoch 27/200
0.5346WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5346
Epoch 28/200
0.5462WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5462
Epoch 29/200
0.5480WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5480
```

```
Epoch 30/200
0.5497WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 50s 887ms/step - loss: 0.6866 -
accuracy: 0.5497
Epoch 31/200
57/57 [===================== ] - ETA: 0s - loss: 0.6891 - accuracy:
0.5702WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5702
Epoch 32/200
0.5488WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5488
Epoch 33/200
57/57 [========================= ] - ETA: 0s - loss: 0.6872 - accuracy:
0.5284WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5284
Epoch 34/200
0.5417WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 47s 823ms/step - loss: 0.6866 -
accuracy: 0.5417
Epoch 35/200
57/57 [========================== ] - ETA: 0s - loss: 0.6831 - accuracy:
0.5560WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5560
Epoch 36/200
57/57 [========================= ] - ETA: 0s - loss: 0.6807 - accuracy:
0.5728WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5728
Epoch 37/200
57/57 [========================= ] - ETA: 0s - loss: 0.6782 - accuracy:
0.5728WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 20s 352ms/step - loss: 0.6782 -
accuracy: 0.5728
Epoch 38/200
57/57 [========================= ] - ETA: 0s - loss: 0.6820 - accuracy:
0.5604WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5604
Epoch 39/200
57/57 [========================== ] - ETA: 0s - loss: 0.6744 - accuracy:
0.5933WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
```

```
is not available. Available metrics are: loss,accuracy
accuracy: 0.5933
Epoch 40/200
0.5933WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5933
Epoch 41/200
57/57 [========================== ] - ETA: 0s - loss: 0.6738 - accuracy:
0.5941WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5941
Epoch 42/200
57/57 [=========================== ] - ETA: 0s - loss: 0.6603 - accuracy:
0.6128WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6128
Epoch 43/200
57/57 [============== ] - ETA: 0s - loss: 0.6719 - accuracy:
0.5737WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss, accuracy
57/57 [============= ] - 22s 392ms/step - loss: 0.6719 -
accuracy: 0.5737
Epoch 44/200
0.5631WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5631
Epoch 45/200
57/57 [============= ] - ETA: 0s - loss: 0.6603 - accuracy:
0.5959WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5959
Epoch 46/200
0.5906WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5906
Epoch 47/200
0.6030WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6030
Epoch 48/200
57/57 [===================== ] - ETA: 0s - loss: 0.6737 - accuracy:
0.5728WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5728
```

```
Epoch 49/200
57/57 [===================== ] - ETA: 0s - loss: 0.6581 - accuracy:
0.5933WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.5933
Epoch 50/200
0.6146WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6146
Epoch 51/200
0.6190WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6190
Epoch 52/200
57/57 [========================= ] - ETA: 0s - loss: 0.6550 - accuracy:
0.6163WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6163
Epoch 53/200
57/57 [========================== ] - ETA: 0s - loss: 0.6353 - accuracy:
0.6297WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 48s 838ms/step - loss: 0.6353 -
accuracy: 0.6297
Epoch 54/200
0.6234WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 50s 875ms/step - loss: 0.6466 -
accuracy: 0.6234
Epoch 55/200
57/57 [========================= ] - ETA: 0s - loss: 0.6411 - accuracy:
0.6261WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6261
Epoch 56/200
0.6217WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6217
Epoch 57/200
57/57 [========================= ] - ETA: 0s - loss: 0.6581 - accuracy:
0.6128WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6128
Epoch 58/200
57/57 [========================== ] - ETA: 0s - loss: 0.6472 - accuracy:
0.6083WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
```

```
is not available. Available metrics are: loss,accuracy
57/57 [================= ] - 51s 899ms/step - loss: 0.6472 -
accuracy: 0.6083
Epoch 59/200
0.6385WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6385
Epoch 60/200
0.6226WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6226
Epoch 61/200
57/57 [========================== ] - ETA: 0s - loss: 0.6261 - accuracy:
0.6456WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6456
Epoch 62/200
57/57 [============== ] - ETA: 0s - loss: 0.6241 - accuracy:
0.6439WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss, accuracy
57/57 [============= ] - 50s 873ms/step - loss: 0.6241 -
accuracy: 0.6439
Epoch 63/200
0.6385WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6385
Epoch 64/200
57/57 [============ ] - ETA: 0s - loss: 0.6261 - accuracy:
0.6554WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6554
Epoch 65/200
0.6750WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6750
Epoch 66/200
0.6448WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6448
Epoch 67/200
57/57 [=========================== ] - ETA: 0s - loss: 0.6000 - accuracy:
0.6696WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6696
```

```
Epoch 68/200
0.6758WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6758
Epoch 69/200
0.6767WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6767
Epoch 70/200
0.6572WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6572
Epoch 71/200
57/57 [========================= ] - ETA: 0s - loss: 0.6036 - accuracy:
0.6643WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6643
Epoch 72/200
57/57 [========================= ] - ETA: 0s - loss: 0.5950 - accuracy:
0.7007WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 22s 391ms/step - loss: 0.5950 -
accuracy: 0.7007
Epoch 73/200
57/57 [========================= ] - ETA: 0s - loss: 0.6010 - accuracy:
0.6767WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 21s 371ms/step - loss: 0.6010 -
accuracy: 0.6767
Epoch 74/200
57/57 [========================= ] - ETA: 0s - loss: 0.5845 - accuracy:
0.6803WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6803
Epoch 75/200
57/57 [========================= ] - ETA: 0s - loss: 0.5599 - accuracy:
0.7114WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 24s 427ms/step - loss: 0.5599 -
accuracy: 0.7114
Epoch 76/200
57/57 [========================= ] - ETA: 0s - loss: 0.5740 - accuracy:
0.7034WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7034
Epoch 77/200
57/57 [========================== ] - ETA: 0s - loss: 0.5890 - accuracy:
0.7034WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
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is not available. Available metrics are: loss,accuracy
accuracy: 0.7034
Epoch 78/200
0.6998WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.6998
Epoch 79/200
57/57 [========================== ] - ETA: 0s - loss: 0.5539 - accuracy:
0.7105WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7105
Epoch 80/200
0.7131WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7131
Epoch 81/200
57/57 [============== ] - ETA: 0s - loss: 0.5720 - accuracy:
0.7025WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss, accuracy
57/57 [============= ] - 26s 458ms/step - loss: 0.5720 -
accuracy: 0.7025
Epoch 82/200
0.7425WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7425
Epoch 83/200
57/57 [=========================== ] - ETA: 0s - loss: 0.5099 - accuracy:
0.7558WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7558
Epoch 84/200
57/57 [========================== ] - ETA: 0s - loss: 0.5500 - accuracy:
0.7176WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7176
Epoch 85/200
0.7380WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7380
Epoch 86/200
57/57 [==================== ] - ETA: 0s - loss: 0.5316 - accuracy:
0.7318WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7318
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Epoch 87/200
57/57 [===================== ] - ETA: 0s - loss: 0.5143 - accuracy:
0.7425WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7425
Epoch 88/200
57/57 [============= ] - ETA: 0s - loss: 0.4957 - accuracy:
0.7593WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7593
Epoch 89/200
0.7664WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7664
Epoch 90/200
57/57 [========================= ] - ETA: 0s - loss: 0.4879 - accuracy:
0.7638WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7638
Epoch 91/200
0.7647WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 25s 443ms/step - loss: 0.4889 -
accuracy: 0.7647
Epoch 92/200
0.7789WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 25s 445ms/step - loss: 0.4656 -
accuracy: 0.7789
Epoch 93/200
57/57 [========================= ] - ETA: 0s - loss: 0.4707 - accuracy:
0.7718WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7718
Epoch 94/200
0.8108WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 24s 424ms/step - loss: 0.4484 -
accuracy: 0.8108
Epoch 95/200
57/57 [========================= ] - ETA: 0s - loss: 0.4595 - accuracy:
0.7869WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7869
Epoch 96/200
0.7718WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
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is not available. Available metrics are: loss,accuracy
accuracy: 0.7718
Epoch 97/200
0.7957WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.7957
Epoch 98/200
57/57 [========================== ] - ETA: 0s - loss: 0.4308 - accuracy:
0.8020WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8020
Epoch 99/200
0.8099WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8099
Epoch 100/200
57/57 [============== ] - ETA: 0s - loss: 0.4048 - accuracy:
0.8046WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss, accuracy
57/57 [============ ] - 24s 425ms/step - loss: 0.4048 -
accuracy: 0.8046
Epoch 101/200
0.8268WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8268
Epoch 102/200
57/57 [============ ] - ETA: 0s - loss: 0.4237 - accuracy:
0.8046WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8046
Epoch 103/200
57/57 [========================= ] - ETA: 0s - loss: 0.4039 - accuracy:
0.8179WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8179
Epoch 104/200
0.8348WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8348
Epoch 105/200
57/57 [===================== ] - ETA: 0s - loss: 0.4013 - accuracy:
0.8126WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8126
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Epoch 106/200
0.8206WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 27s 467ms/step - loss: 0.3892 -
accuracy: 0.8206
Epoch 107/200
57/57 [=========================== ] - ETA: 0s - loss: 0.3991 - accuracy:
0.8268WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8268
Epoch 108/200
0.8321WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8321
Epoch 109/200
57/57 [========================= ] - ETA: 0s - loss: 0.3449 - accuracy:
0.8517WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8517
Epoch 110/200
0.8321WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 28s 486ms/step - loss: 0.3660 -
accuracy: 0.8321
Epoch 111/200
57/57 [=========================== ] - ETA: 0s - loss: 0.3239 - accuracy:
0.8526WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 26s 457ms/step - loss: 0.3239 -
accuracy: 0.8526
Epoch 112/200
57/57 [========================= ] - ETA: 0s - loss: 0.3350 - accuracy:
0.8552WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8552
Epoch 113/200
57/57 [========================= ] - ETA: 0s - loss: 0.3270 - accuracy:
0.8632WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 25s 430ms/step - loss: 0.3270 -
accuracy: 0.8632
Epoch 114/200
57/57 [========================== ] - ETA: 0s - loss: 0.3674 - accuracy:
0.8393WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8393
Epoch 115/200
57/57 [========================== ] - ETA: 0s - loss: 0.3181 - accuracy:
0.8597WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
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is not available. Available metrics are: loss,accuracy
accuracy: 0.8597
Epoch 116/200
0.8375WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8375
Epoch 117/200
57/57 [========================== ] - ETA: 0s - loss: 0.3192 - accuracy:
0.8686WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8686
Epoch 118/200
57/57 [========================= ] - ETA: 0s - loss: 0.3201 - accuracy:
0.8615WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8615
Epoch 119/200
57/57 [============== ] - ETA: 0s - loss: 0.2874 - accuracy:
0.8694WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss, accuracy
57/57 [============= ] - 25s 430ms/step - loss: 0.2874 -
accuracy: 0.8694
Epoch 120/200
0.8783WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8783
Epoch 121/200
0.8757WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8757
Epoch 122/200
57/57 [========================== ] - ETA: 0s - loss: 0.2736 - accuracy:
0.8828WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8828
Epoch 123/200
57/57 [=========================== ] - ETA: 0s - loss: 0.3097 - accuracy:
0.8766WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8766
Epoch 124/200
0.8828WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8828
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Epoch 125/200
0.8863WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8863
Epoch 126/200
0.8917WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8917
Epoch 127/200
0.8917WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8917
Epoch 128/200
57/57 [========================= ] - ETA: 0s - loss: 0.2908 - accuracy:
0.8721WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8721
Epoch 129/200
57/57 [========================== ] - ETA: 0s - loss: 0.2453 - accuracy:
0.8979WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 23s 393ms/step - loss: 0.2453 -
accuracy: 0.8979
Epoch 130/200
57/57 [========================= ] - ETA: 0s - loss: 0.2597 - accuracy:
0.8925WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 22s 390ms/step - loss: 0.2597 -
accuracy: 0.8925
Epoch 131/200
57/57 [========================== ] - ETA: 0s - loss: 0.2472 - accuracy:
0.9076WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9076
Epoch 132/200
0.8996WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 24s 429ms/step - loss: 0.2694 -
accuracy: 0.8996
Epoch 133/200
57/57 [========================== ] - ETA: 0s - loss: 0.2662 - accuracy:
0.8899WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8899
Epoch 134/200
57/57 [========================= ] - ETA: 0s - loss: 0.3109 - accuracy:
0.8792WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
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is not available. Available metrics are: loss,accuracy
57/57 [================= ] - 24s 425ms/step - loss: 0.3109 -
accuracy: 0.8792
Epoch 135/200
57/57 [========================= ] - ETA: 0s - loss: 0.2409 - accuracy:
0.9076WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9076
Epoch 136/200
57/57 [========================== ] - ETA: 0s - loss: 0.2320 - accuracy:
0.9032WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9032
Epoch 137/200
57/57 [========================= ] - ETA: 0s - loss: 0.2225 - accuracy:
0.9156WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9156
Epoch 138/200
57/57 [============== ] - ETA: 0s - loss: 0.2412 - accuracy:
0.9041WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss, accuracy
57/57 [============= ] - 26s 452ms/step - loss: 0.2412 -
accuracy: 0.9041
Epoch 139/200
0.9130WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9130
Epoch 140/200
57/57 [============= ] - ETA: 0s - loss: 0.2575 - accuracy:
0.9059WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9059
Epoch 141/200
57/57 [========================= ] - ETA: 0s - loss: 0.2278 - accuracy:
0.9094WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9094
Epoch 142/200
57/57 [=========================== ] - ETA: 0s - loss: 0.1992 - accuracy:
0.9325WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9325
Epoch 143/200
57/57 [=========================== ] - ETA: 0s - loss: 0.2151 - accuracy:
0.9147WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9147
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Epoch 144/200
0.9076WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9076
Epoch 145/200
57/57 [=========================== ] - ETA: 0s - loss: 0.1787 - accuracy:
0.9298WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9298
Epoch 146/200
0.9174WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9174
Epoch 147/200
57/57 [========================= ] - ETA: 0s - loss: 0.1938 - accuracy:
0.9201WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9201
Epoch 148/200
0.9139WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 20s 343ms/step - loss: 0.2095 -
accuracy: 0.9139
Epoch 149/200
57/57 [========================= ] - ETA: 0s - loss: 0.1701 - accuracy:
0.9352WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 20s 358ms/step - loss: 0.1701 -
accuracy: 0.9352
Epoch 150/200
57/57 [========================= ] - ETA: 0s - loss: 0.2032 - accuracy:
0.9245WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9245
Epoch 151/200
57/57 [========================= ] - ETA: 0s - loss: 0.1895 - accuracy:
0.9218WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============= ] - 20s 354ms/step - loss: 0.1895 -
accuracy: 0.9218
Epoch 152/200
57/57 [========================= ] - ETA: 0s - loss: 0.1949 - accuracy:
0.9210WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9210
Epoch 153/200
57/57 [========================== ] - ETA: 0s - loss: 0.1761 - accuracy:
0.9325WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
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is not available. Available metrics are: loss,accuracy
accuracy: 0.9325
Epoch 154/200
57/57 [========================= ] - ETA: 0s - loss: 0.2756 - accuracy:
0.8988WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.8988
Epoch 155/200
57/57 [========================== ] - ETA: 0s - loss: 0.2010 - accuracy:
0.9227WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9227
Epoch 156/200
57/57 [===================== ] - ETA: 0s - loss: 0.2119 - accuracy:
0.9165WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9165
Epoch 157/200
57/57 [============== ] - ETA: 0s - loss: 0.1721 - accuracy:
0.9361WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss, accuracy
57/57 [============= ] - 21s 367ms/step - loss: 0.1721 -
accuracy: 0.9361
Epoch 158/200
0.9254WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9254
Epoch 159/200
57/57 [============ ] - ETA: 0s - loss: 0.1802 - accuracy:
0.9343WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9343
Epoch 160/200
57/57 [========================== ] - ETA: 0s - loss: 0.1783 - accuracy:
0.9352WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9352
Epoch 161/200
0.9405WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9405
Epoch 162/200
57/57 [==================== ] - ETA: 0s - loss: 0.1316 - accuracy:
0.9449WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9449
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Epoch 163/200
0.9245WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9245
Epoch 164/200
0.9405WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9405
Epoch 165/200
0.9352WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9352
Epoch 166/200
57/57 [========================= ] - ETA: 0s - loss: 0.1590 - accuracy:
0.9449WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9449
Epoch 167/200
0.9458WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 22s 386ms/step - loss: 0.1467 -
accuracy: 0.9458
Epoch 168/200
57/57 [========================== ] - ETA: 0s - loss: 0.1885 - accuracy:
0.9334WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 22s 393ms/step - loss: 0.1885 -
accuracy: 0.9334
Epoch 169/200
57/57 [========================== ] - ETA: 0s - loss: 0.1854 - accuracy:
0.9307WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9307
Epoch 170/200
57/57 [========================= ] - ETA: 0s - loss: 0.1721 - accuracy:
0.9307WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 22s 393ms/step - loss: 0.1721 -
accuracy: 0.9307
Epoch 171/200
57/57 [========================= ] - ETA: 0s - loss: 0.1529 - accuracy:
0.9325WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9325
Epoch 172/200
0.9503WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
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is not available. Available metrics are: loss,accuracy
accuracy: 0.9503
Epoch 173/200
57/57 [========================== ] - ETA: 0s - loss: 0.1168 - accuracy:
0.9609WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9609
Epoch 174/200
0.9369WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9369
Epoch 175/200
0.9440WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9440
Epoch 176/200
57/57 [============== ] - ETA: 0s - loss: 0.1286 - accuracy:
0.9520WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss, accuracy
57/57 [============= ] - 21s 372ms/step - loss: 0.1286 -
accuracy: 0.9520
Epoch 177/200
0.9334WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9334
Epoch 178/200
57/57 [============ ] - ETA: 0s - loss: 0.1302 - accuracy:
0.9512WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9512
Epoch 179/200
57/57 [=========================== ] - ETA: 0s - loss: 0.1262 - accuracy:
0.9467WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9467
Epoch 180/200
0.9272WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9272
Epoch 181/200
0.9263WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9263
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Epoch 182/200
0.9467WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9467
Epoch 183/200
57/57 [==================== ] - ETA: 0s - loss: 0.1202 - accuracy:
0.9600WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9600
Epoch 184/200
0.9520WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9520
Epoch 185/200
57/57 [========================= ] - ETA: 0s - loss: 0.1202 - accuracy:
0.9512WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9512
Epoch 186/200
0.9556 WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 3025s 54s/step - loss: 0.1349 -
accuracy: 0.9556
Epoch 187/200
57/57 [========================= ] - ETA: 0s - loss: 0.1997 - accuracy:
0.9263WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 22s 389ms/step - loss: 0.1997 -
accuracy: 0.9263
Epoch 188/200
57/57 [========================= ] - ETA: 0s - loss: 0.1505 - accuracy:
0.9405WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9405
Epoch 189/200
57/57 [========================= ] - ETA: 0s - loss: 0.1039 - accuracy:
0.9654WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 26s 461ms/step - loss: 0.1039 -
accuracy: 0.9654
Epoch 190/200
0.9414WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9414
Epoch 191/200
57/57 [========================== ] - ETA: 0s - loss: 0.1330 - accuracy:
0.9432WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
```

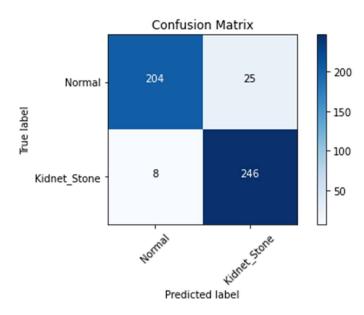
```
is not available. Available metrics are: loss,accuracy
accuracy: 0.9432
Epoch 192/200
0.9467WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9467
Epoch 193/200
57/57 [========================== ] - ETA: 0s - loss: 0.1188 - accuracy:
0.9583WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9583
Epoch 194/200
57/57 [===================== ] - ETA: 0s - loss: 0.1377 - accuracy:
0.9512WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9512
Epoch 195/200
57/57 [============== ] - ETA: 0s - loss: 0.1219 - accuracy:
0.9618WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss, accuracy
57/57 [============ ] - 49s 855ms/step - loss: 0.1219 -
accuracy: 0.9618
Epoch 196/200
0.9529WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9529
Epoch 197/200
57/57 [============ ] - ETA: 0s - loss: 0.1485 - accuracy:
0.9547WARNING:tensorflow:Early stopping conditioned on metric `val loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9547
Epoch 198/200
57/57 [========================= ] - ETA: 0s - loss: 0.1002 - accuracy:
0.9654WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============== ] - 45s 793ms/step - loss: 0.1002 -
accuracy: 0.9654
Epoch 199/200
0.9440WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
accuracy: 0.9440
Epoch 200/200
57/57 [===================== ] - ETA: 0s - loss: 0.1066 - accuracy:
0.9574WARNING:tensorflow:Early stopping conditioned on metric `val_loss` which
is not available. Available metrics are: loss,accuracy
57/57 [============================ ] - 49s 867ms/step - loss: 0.1066 -
accuracy: 0.9574
```

Accuracy: 93.3747410774231

plt.title(title)

```
SVM CLASSIFIER
                                                                             In [52]:
Xtrain.shape
                                                                            Out[52]:
(1126, 128, 128, 3)
                                                                             In [53]:
Xtest.shape
                                                                            Out[53]:
(483, 128, 128, 3)
                                                                             In [54]:
Xtrain_SVC = Xtrain.reshape(1126, 128 * 128 * 3)
Xtest_SVC = Xtest.reshape(483, 128 * 128 * 3)
                                                                             In [55]:
model_SVC = SVC(C = 1,kernel='poly',gamma = 'auto')
model_SVC.fit(Xtrain_SVC, Ytrain)
                                                                            Out[55]:
SVC(C=1, gamma='auto', kernel='poly')
                                                                             In [56]:
prediction = model_SVC.predict(Xtest_SVC)
accuracy = model_SVC.score(Xtest_SVC , Ytest)
                                                                             In [57]:
train_accuracy = model_SVC.score(Xtrain_SVC, Ytrain)
print("Train_Accuracy", train_accuracy * 100)
Train Accuracy 100.0
                                                                             In [58]:
print("Test_Accuracy ",accuracy * 100)
Test_Accuracy 93.16770186335404
 Confusion Matrix
                                                                             In [59]:
# cm = confusion matrix(y true = Ytest, y pred = np.argmax(predictions,axis=-
1))
cm = confusion_matrix(Ytest , (prediction > 0.75) * 1)
                                                                             In [60]:
def plot confusion matrix(cm, classes,
                           normalize=False,
                           title='Confusion matrix',
                           cmap=plt.cm.Blues):
    This function prints and plots the confusion matrix.
    Normalization can be applied by setting `normalize=True`.
    plt.imshow(cm, interpolation='nearest', cmap=cmap)
```

```
plt.colorbar()
    tick marks = np.arange(len(classes))
    plt.xticks(tick_marks, classes, rotation=45)
    plt.yticks(tick_marks, classes)
    if normalize:
        cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
        print("Normalized confusion matrix")
    else:
        print('Confusion matrix, without normalization')
    print(cm)
    thresh = cm.max() / 2.
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
        plt.text(j, i, cm[i, j],
            horizontalalignment="center",
            color="white" if cm[i, j] > thresh else "black")
    plt.tight_layout()
    plt.ylabel('True label')
    plt.xlabel('Predicted label')
                                                                           In [61]:
cm_plot_labels = ['Normal','Kidnet_Stone']
plot_confusion_matrix(cm = cm, classes = cm_plot_labels, title = 'Confusion'
Matrix')
Confusion matrix, without normalization
[[204 25]
 [ 8 246]]
```



In [62]:

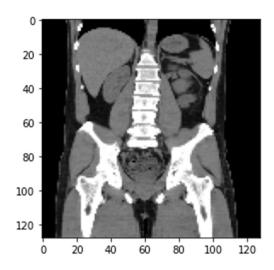
```
def image_prediction(img):
    plt.imshow(img)
    img1=img.reshape(1,128,128,3)
    predict = model3.predict(img1)
    if ((predict > 0.75) * 1):
```

```
print("The condition is normal and stable.")
    else:
        print("Person is having kidney stone(s)")
                                                                         In [63]:
image_prediction(Xtest[150])
1/1 [=======] - 0s 32ms/step
Person is having kidney stone(s)
 20
 40
 60
 80
 100
120
        20
            40
                60
                     80
                         100
    0
                             120
                                                                         In [64]:
image_prediction(Xtest[50])
1/1 [======] - 0s 34ms/step
The condition is normal and stable.
 20
 40
 60
 80
100
120
            40
                60
                     80
                         100
                             120
                                                                          In []:
!pip install split-folders
                                                                         In [74]:
def image_prediction_KMeans(img):
    img = img.reshape(128,128,3)
    plt.imshow(img)
    img1 = img.reshape(1, 128 * 128 * 3)
```

predict = kmeans_model.predict(img1)

```
print(predict)
if (predict):
    print("Person is having kidney stone(s)")
else:
    print("The condition is normal and stable.")

In [77]:
path_2 = "C:/Users/prajwal mr/Downloads/archive/CT_SCAN/Normal/N1.png"
image_array_2 = cv2.imread(path_2)
new_image_2 = cv2.resize(image_array_2,(128,128))
plt.imshow(new_image_2)
plt.show()
```



model3.save('kidney_stones_model.h5')

In [95]:

In []:

4. Conclusion

In conclusion, a kidney stone detection project utilizing digital image processing (DIP) techniques can be an interesting research endeavor. However, it is important to note that DIP-based kidney stone detection is not commonly used in clinical settings for diagnosing kidney stones. Medical imaging techniques such as ultrasound, X-ray, or computed tomography (CT) scans are the established methods for kidney stone detection and diagnosis

Ultimately, the aim of a kidney stone detection project using DIP would be to explore the feasibility of utilizing image processing techniques to aid in the identification and diagnosis of kidney stones. While it may not replace established medical imaging methods, it could potentially contribute to the development of supplementary tools or assist in early identification of kidney stones, enhancing patient care and treatment.