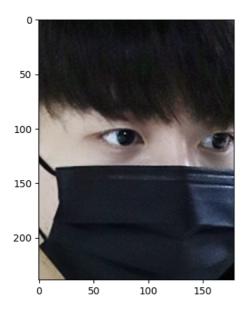
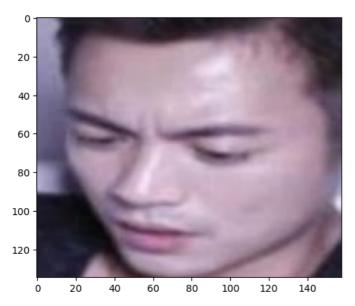
```
!pip install opency-python
     Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>
     Requirement already satisfied: opencv-python in /usr/local/lib/python3.10/dist-packages (4.7.0.72)
     Requirement already satisfied: numpy>=1.19.3 in /usr/local/lib/python3.10/dist-packages (from opency-python) (1.22)
import cv2
import zipfile
with zipfile.ZipFile('/content/finalproject.zip', 'r') as zip_ref:
    zip_ref.extractall('/content/')
import os
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as mpimg
import cv2
from google.colab.patches import cv2_imshow
from PIL import Image
from sklearn.model_selection import train_test_split
with mask files = os.listdir('/content/Final Project/Training Data/with mask')
print(with_mask_files[0:5])
print(with_mask_files[-5:])
     ['with_mask_279.jpg', 'with_mask_162.jpg', 'with_mask_259.jpg', 'with_mask_267.jpg', 'with_mask_28.jpg']
['with_mask_390.jpg', 'with_mask_29.jpg', 'with_mask_361.jpg', 'with_mask_289.jpg']
without_mask_files = os.listdir('/content/Final Project/Training Data/without_mask')
print(without_mask_files[0:5])
print(without_mask_files[-5:])
     ['without_mask_17.jpg', 'without_mask_491.jpg', 'without_mask_96.jpg', 'without_mask_103.jpg', 'without_mask_311.j
['without_mask_319.jpg', 'without_mask_142.jpg', 'without_mask_481.jpg', 'without_mask_43.jpg', 'without_mask_230
print('Number of with mask images:', len(with_mask_files))
print('Number of without mask images:', len(without_mask_files))
     Number of with mask images: 500
     Number of without mask images: 500
# create the labels
with_mask_labels = [1]*500
without_mask_labels = [0]*500
print(with_mask_labels[0:5])
print(without mask labels[0:5])
     [1, 1, 1, 1, 1]
     [0, 0, 0, 0, 0]
print(len(with_mask_labels))
print(len(without_mask_labels))
     500
     500
labels = with_mask_labels + without_mask_labels
print(len(labels))
print(labels[0:5])
print(labels[-5:])
     1000
     [1, 1, 1, 1, 1]
     [0, 0, 0, 0, 0]
```

```
# displaying with mask image
img = mpimg.imread('/content/Final Project/Training Data/with_mask/with_mask_390.jpg')
imgplot = plt.imshow(img)
plt.show()
```



displaying without mask image
img = mpimg.imread('/content/Final Project/Training Data/without_mask/without_mask_319.jpg')
imgplot = plt.imshow(img)
plt.show()



```
# convert images to numpy arrays+
with_mask_path = '/content/Final Project/Training Data/with_mask/'
data = []
for img_file in with_mask_files:
   image = Image.open(with_mask_path + img_file)
   image = image.resize((128,128))
   image = image.convert('RGB')
   image = np.array(image)
   data.append(image)
```

without_mask_path = '/content/Final Project/Training Data/without_mask/'
for img_file in without_mask_files:

```
image = Image.open(without_mask_path + img_file)
  image = image.resize((128,128))
  image = image.convert('RGB')
  image = np.array(image)
  data.append(image)
type(data)
     list
len(data)
     1000
data[0]
    [ 54, 40, 37],
              [ 57, 43, 40],
[ 62, 48, 45]],
             [[ 65, 51, 50],
              [ 65, 51, 50],
[ 63, 49, 48],
              [ 52, 38, 35],
              [ 49, 35, 32],
[ 50, 36, 33]],
             [[ 68, 54, 53],
             [ 64, 50, 49],
[ 56, 42, 41],
              [ 60, 46, 44],
[ 54, 40, 38],
              [ 49, 35, 33]],
             ...,
             [[185, 192, 198],
             [185, 194, 199],
              [208, 218, 225],
              [192, 164, 152],
             [191, 161, 150],
[192, 160, 149]],
             [[218, 225, 231],
             [173, 180, 186],
              [163, 172, 178],
              [192, 164, 152],
             [192, 162, 151],
[193, 161, 150]],
             [[232, 239, 245],
             [221, 228, 234],
              [192, 201, 207],
              [193, 165, 153],
              [193, 163, 152],
[193, 162, 151]]], dtype=uint8)
type(data[0])
     numpy.ndarray
data[0].shape
     (128, 128, 3)
# converting image list and label list to numpy arrays
X = np.array(data)
Y = np.array(labels)
```

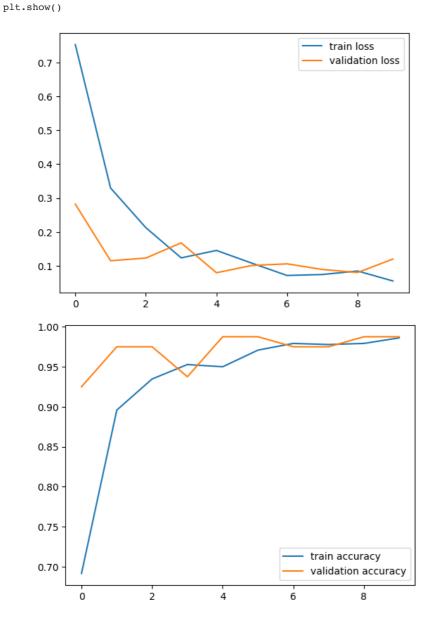
```
type(X)
 numpy.ndarray
type(Y)
 numpy.ndarray
print(X.shape)
print(Y.shape)
 (1000, 128, 128, 3)
 (1000.)
print(Y)
 01
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, random_state=2)
print(X.shape, X_train.shape, X_test.shape)
 (1000, 128, 128, 3) (800, 128, 128, 3) (200, 128, 128, 3)
# scaling the data
X_train_scaled = X_train/255
X test scaled = X test/255
X_train[0]
 array([[[ 82, 73, 58], [ 82, 73, 58], [ 82, 73, 58],
    86,
   [ 99,
     691,
   [103,
    90,
     71],
   [103, 90, 71]],
  [[ 81, 73,
     58],
   [ 79, 70,
     55],
    70,
   [ 79,
     55],
    87,
     701,
   ſ100.
   [103,
    90,
     73],
   [102, 89, 71]],
  [[ 81, 72,
     571,
   [ 79, 70,
     55],
   [ 79,
    70,
     55],
```

```
[104, 89, 73],
             [103, 88, 72],
[103, 88, 72]],
            . . . ,
            [[ 35, 21, 14], [ 34, 19, 12],
             [ 36, 21, 13],
                          3],
             [ 24,
                   10,
             [ 41, 28, 20],
             [ 45, 34, 24]],
            [[ 34, 20, 13],
                          9],
             [ 32, 16,
             [ 38, 20, 13],
             [ 30, 18, 10],
             [ 39, 28, 18],
             [ 42, 32, 22]],
            [[ 34, 19, 12],
             [ 33, 16, 9],
[ 35, 17, 10],
             [ 33,
                   23, 13],
             [ 41, 31, 21],
[ 47, 37, 27]]], dtype=uint8)
X_train_scaled[0]
     array([[[0.32156863, 0.28627451, 0.22745098],
             [0.32156863, 0.28627451, 0.22745098],
             [0.32156863, 0.28627451, 0.22745098],
             [0.38823529, 0.3372549 , 0.27058824],
             [0.40392157, 0.35294118, 0.27843137],
             [0.40392157, 0.35294118, 0.27843137]],
            [[0.31764706, 0.28627451, 0.22745098],
             [0.30980392, 0.2745098 , 0.21568627],
             [0.30980392, 0.2745098, 0.21568627],
             [0.39215686, 0.34117647, 0.2745098],
             [0.40392157, 0.35294118, 0.28627451],
                       , 0.34901961, 0.27843137]],
            [[0.31764706, 0.28235294, 0.22352941],
             [0.30980392, 0.2745098 , 0.21568627],
             [0.30980392, 0.2745098 , 0.21568627],
             [0.40784314, 0.34901961, 0.28627451],
             [0.40392157, 0.34509804, 0.28235294],
             [0.40392157, 0.34509804, 0.28235294]],
            ...,
            [[0.1372549 , 0.08235294, 0.05490196],
             [0.13333333, 0.0745098 , 0.04705882],
             [0.14117647, 0.08235294, 0.05098039],
             [0.09411765, 0.03921569, 0.01176471],
             [0.16078431, 0.10980392, 0.07843137],
             [0.17647059, 0.13333333, 0.09411765]],
            [[0.13333333, 0.07843137, 0.05098039],
             [0.1254902 , 0.0627451 , 0.03529412], [0.14901961, 0.07843137, 0.05098039],
             [0.11764706, 0.07058824, 0.03921569],
             [0.15294118, 0.10980392, 0.07058824],
             [0.16470588, 0.1254902 , 0.08627451]],
            [[0.13333333, 0.0745098 , 0.04705882],
             [0.12941176, 0.0627451 , 0.03529412],
             [0.1372549 , 0.06666667, 0.03921569],
             [0.12941176, 0.09019608, 0.05098039],
             [0.16078431, 0.12156863, 0.08235294],
             [0.18431373, 0.14509804, 0.10588235]]])
```

```
import tensorflow as tf
from tensorflow import keras
num of classes = 2
model = keras.Sequential()
model.add(keras.layers.Conv2D(32, kernel_size=(3,3), activation='relu', input_shape=(128,128,3)))
model.add(keras.layers.MaxPooling2D(pool_size=(2,2)))
model.add(keras.layers.Conv2D(64, kernel_size=(3,3), activation='relu'))
model.add(keras.layers.MaxPooling2D(pool_size=(2,2)))
model.add(keras.layers.Flatten())
model.add(keras.layers.Dense(128, activation='relu'))
model.add(keras.layers.Dropout(0.5))
model.add(keras.layers.Dense(64, activation='relu'))
model.add(keras.layers.Dropout(0.5))
model.add(keras.layers.Dense(num_of_classes, activation='sigmoid'))
# compile the neural network
model.compile(optimizer='adam',
           loss='sparse_categorical_crossentropy',
           metrics=['acc'])
# training the neural network
history = model.fit(X train scaled, Y train, validation split=0.1, epochs=10)
    Epoch 1/10
   23/23 [===========] - 24s 911ms/step - loss: 0.7523 - acc: 0.6917 - val loss: 0.2820 - val acc
    Epoch 2/10
    23/23 [====
                   ========== ] - 22s 951ms/step - loss: 0.3294 - acc: 0.8958 - val loss: 0.1151 - val acc
    Epoch 3/10
   23/23 [============ ] - 20s 889ms/step - loss: 0.2124 - acc: 0.9347 - val loss: 0.1231 - val acc
   Epoch 4/10
    23/23 [============== ] - 21s 939ms/step - loss: 0.1236 - acc: 0.9528 - val_loss: 0.1678 - val_acc
    Epoch 5/10
                  23/23 [=====
   Epoch 6/10
              23/23 [=====
   Epoch 7/10
   23/23 [============ ] - 22s 943ms/step - loss: 0.0717 - acc: 0.9792 - val_loss: 0.1059 - val_acc
   Epoch 8/10
                    ==========] - 21s 900ms/step - loss: 0.0745 - acc: 0.9778 - val_loss: 0.0894 - val_acc
    23/23 [=====
   Epoch 9/10
   23/23 [============ ] - 22s 952ms/step - loss: 0.0848 - acc: 0.9792 - val loss: 0.0803 - val acc
   Epoch 10/10
    Calculating Model Accuracy
#Model evaluation
loss, accuracy = model.evaluate(X_test_scaled, Y_test)
print('Test Accuracy =', accuracy)
   7/7 [=========] - 1s 197ms/step - loss: 0.0271 - acc: 0.9850
   Test Accuracy = 0.9850000143051147
Plotting the graph
h = history
# plot the loss value
plt.plot(h.history['loss'], label='train loss')
plt.plot(h.history['val_loss'], label='validation loss')
plt.legend()
```

```
plt.show()

# plot the accuracy value
plt.plot(h.history['acc'], label='train accuracy')
plt.plot(h.history['val_acc'], label='validation accuracy')
plt.legend()
```



Predicting with Mask

```
input_image_path = input('Path of the image to be predicted: ')
input_image = cv2.imread(input_image_path)
cv2_imshow(input_image)
input_image_resized = cv2.resize(input_image, (128,128))
input_image_scaled = input_image_resized/255
input_image_reshaped = np.reshape(input_image_scaled, [1,128,128,3])
input_prediction = model.predict(input_image_reshaped)
print(input_prediction)

input_pred_label = np.argmax(input_prediction)

print(type(input_pred_label))
if input_pred_label == 1:
```

```
print('The person in the image is wearing a mask')
else:
    print('The person in the image is not wearing a mask')
```

Path of the image to be predicted: /content/anshumask.jpeg

```
1/1 [======] - 0s 48ms/step [[0.19748174 0.9925925 ]] <class 'numpy.int64'>
The person in the image is wearing a mask
```

Predicting No Mask

```
#code begins from here
input_image_path = input('Path of the image to be predicted: ')
input_image = cv2.imread(input_image_path)
cv2_imshow(input_image)
input_image_resized = cv2.resize(input_image, (128,128))
input_image_scaled = input_image_resized/255
input_image_reshaped = np.reshape(input_image_scaled, [1,128,128,3])
input_prediction = model.predict(input_image_reshaped)
print(input_prediction)

input_pred_label = np.argmax(input_prediction)

print(type(input_pred_label))
if input_pred_label == 0:
    print('The person in the image is wearing a mask')
else:
    print('The person in the image is not wearing a mask')
```

Path of the image to be predicted: /content/WhatsApp Image 2023-05-09 at 11.12.05 PM (1).jpg



1/1 [======] - 0s 50ms/step [[0.3353671 0.8706998]] <class 'numpy.int64'> The person in the image is not wearing a mask