

Facemask Detector in Surveillance for COVID-19

Sai Vignesh Ramisetty, D Madhumita, K Yashwanth Chowdary



Abstract: Due to this unexpected pandemic we are going on these days, wearing a face mask became mandatory to save ourselves as well as others from the virus. But it is difficult to monitor every citizen whether he is wearing a mask or not. But it is very important. So, to overcome this problem we came up with a solution to monitor every citizen using a deep learning concept. So, we are developing a face mask detector with opencv/keras. This helps us to easily identify the persons wearing masks or not which helps us in taking safety measures according to it. We tried using different types of platforms such as mobilev2net and resnet architecture but the accuracy of resnet architecture is more compared to the other architecture.

Keywords: This helps us to easily identify the persons wearing masks or not which helps us in taking safety measures according to it.

I. INTRODUCTION

Developing this face mask detector is not only the way but we should develop it in a more portable way which can be used in any area. So, to make it portable we are using a Resnet classifier. This is the only way to make this project portable and user friendly. Basically, there are two phases in this face mask detector. Phase1: Train the face mask detector. Phase2: Apply face mask detector. Phase 1 is the basic step in which we will train the project using the datasets and train it to vary between different types of images. This will be the testing and comparison set to the real images. In phase 2, the given image is compared with the stored dataset and gives us the original output whether the person is wearing a mask or not. The detailed explanation is in phase 2 the face mask classifier is loaded from the disk, and then the camera detects the image and extracts the ROI of each face. This ROI is applied to each face mask classifier to detect the mask. Finally, it gives the output.

II. OBJECTIVE

We aim to design a hands-free entry system using a face mask detector in surveillance to combat the further spread of the virus.

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This will ensure to reduce the transmission of pathogens on high-touch surfaces, like door handles, and to prevent entry in a community area without a face mask. To accomplish this task, we will be fine-tuning the Resnet architecture, which will help us in training hundreds of layers quickly and making sure that there won't be a drop in the training percentage.

III. LITERATURE REVIEW

[1] The vision towards computers is changing every 13 months. During this change using the computer technology in the neuro-scientific field also came into light in which this face detection is also one of the parts. [2] Algorithms based on principle component analysis are the bases of the numerous studies in neuroscience and algorithmic face detection literature. Eigenvector is one of the main concepts in differentiating and grouping the data to train the device. [3] Every face is treated as a two-dimensional face rather than three dimensional to make the identifying easier. But these images can be identified in any colour but the image should not be tinted with green colour. [4] A view-based multiple-observer eigenspace technique is proposed for use in face recognition under variable pose. In addition, a modular eigenspace description technique is used which incorporates salient features such as the eyes, nose and mouth, in an eigen feature layer. This modular representation yields higher recognition rates as well as a more robust framework for face recognition. An automatic feature extraction technique using feature Eigen templates. [5] For face detection, Haar-Cascades were used and for face recognition Eigenfaces, Fisherfaces and Local binary pattern histograms were used. The difference between face detection and identification is, face detection is to identify a face from an image and locate the face. Face recognition is making the decision "whose face is it?", using an image database. The technologies available in the Open-Computer-Vision (OpenCV) library and methodology to implement them using Python.[6] Human face localization and detection is often the first step in applications such as video surveillance, human computer interface, face recognition and image database management. Locating and tracking human faces is a prerequisite for face recognition and/or facial expressions analysis, although it is often assumed that a normalized face image is available. In this paper we intend to implement the Haar-Classifer for Face detection and tracking based on the HaarFeatures. This is done using Haar Classifier through Raspberry Pi BCM2835 CPU processor which is a combination of SoC with GPU based Architecture. SimpleCV and OpenCV libraries are used for face detection and tracking the head poses position.



IV. SOFTWARE REQUIREMENTS

Tensorflow & Keras: Tensorflow is used to achieve following

- Data augmentation
- Loading the MobilNetV2 classifier
- Building a new fully-connected (FC) head
- Pre-processing
- Loading image data

Sklearn: sklearn is used for binarizing class labels, segmenting our dataset, and printing a classification report.

Imutils: This helps in listing the paths of images inside the dataset.

V. PROPOSED SYSTEM

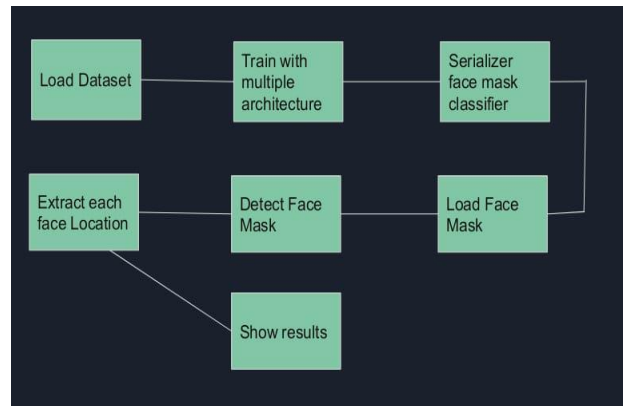
We will take an input from Real-time video as an input and this video is processed using the algorithm developed by us. We use the testing algorithm to detect if the person identified in the image is wearing a mask or not this then gives a digital output which includes the status of the person and the accuracy of the output. This output will be displayed on the screen.

VI. WORKING

A dataset containing images will be used to train and test the code. This dataset contains 690 images with people wearing masks and 600 images with people without a mask . 80% of the dataset is used for training the algorithm and the remaining 20% is used for testing. In the training at first, we will be loading the complete dataset of images as a list. We will be using two arrays in which one of the arrays will contain the data(images) and the other one contains the label(names) of the image. Then both arrays are changed into a numpy array. Now we split the data set into training and testing parts. We have to arrange every image in the dataset according to our requirements so we will be using different syntaxes to alter the dimensions of every image now using the Resnet architecture which was designed to enable hundreds or thousands of convolutional layers. Here we'll focus on loading our face mask detection dataset from disk, training a model (using Keras/TensorFlow) on this dataset, and then serializing the face mask detector to disk. Using this resnet we will create a base model which is used to train the algorithm. This base model is used in prediction of different images in the dataset and the label which is used in the second array is matched with these images. This is going to help in serializing the data so this will end the training part of the system. Coming to the testing part a Real-time video will be given as an input. A video is nothing but a combination of frames so in this testing part we will extract frame from this Real-time video every 1 second this frame is going to be used for mask detection once the frame is extracted the algorithm will identify the person in the image and bounds the face of the person in a rectangular box. When an frame is extracted it is converted into data which is then stored into an array if the value of the array happens to be greater than 0 it means that the face of the person is identified by the algorithm. According to the output if the person is wearing a mask he will be bound with a green box. If not he will be bound with a red box every image will be compared with a set of images in the data set and the output is given according to the results.

So, the value stored in the array will help us identify the face of the person in the image as soon as this happens the detection of the mask takes place so if the person is wearing a mask it's going to print yes along with its accuracy percentage.

VII. BLOCK DIAGRAM

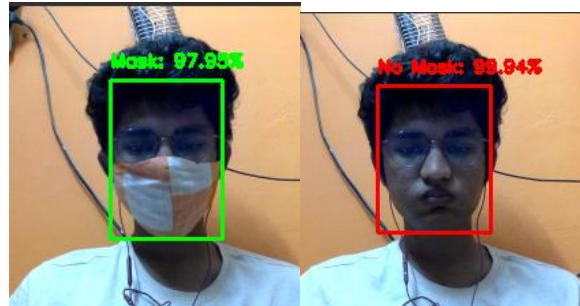


VIII. RESULTS

We have got an accuracy of 89%. With 1:4 ratio of testing: training data.

	precision
with mask	0.89
without mask	0.90
accuracy	0.89

These are our simulations for the same:



IX. CONCLUSION

This paper concludes the development and working of face mask detector using mobile v2 classifier and this is very portable and the accuracy and precision are much better than compared to other systems. By further developing this concept we can add this to the entrance which automatically opens the door on detecting the mask.

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