

AI Face Mask Detection System: A Comprehensive Analysis and Implementation

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Abstract: In the face of the ongoing global pandemic, wearing face masks has become a crucial preventive measure. To ensure public safety, there is a pressing need for monitoring of mask compliance in public spaces. This paper presents a detailed exploration and implementation of an Artificial Intelligence (AI) based Face Mask Detection System. The proposed system leverages state-of-the-art computer vision techniques and deep learning algorithms to accurately identify individuals wearing masks and those without masks. The system architecture involves several key components. First, an image acquisition module captures real-time video from surveillance cameras. Next, a pre-processing stage enhances image quality and reduces noise, ensuring necessary input for the AI model. The heart of the system lies in the Convolutional Neural Network (CNN) model, trained on an extensive dataset comprising diverse facial images with and without masks. Transfer learning techniques are employed, utilizing pre-trained models such as ResNet and MobileNet, to enhance the efficiency of mask detection.

Keywords: Artificial Intelligence, Face Mask Detection, Computer Vision, Deep Learning, Convolutional Neural Networks

I. INTRODUCTION

Face mask detection using artificial intelligence (AI) is a technology that utilizes computer vision and machine learning techniques to automatically identify whether a person is wearing a face mask or not. This technology has gained significant importance, especially during the COVID-19 pandemic, as wearing masks in public spaces has been recommended as an effective measure to prevent the spread of the virus. AI face mask detection systems have been deployed in various public areas, including, hospitals, schools, and retail stores, to ensure compliance with mask-wearing guidelines.

Components of AI Face Mask Detection:

Computer Vision:

- **Face Detection:** Computer vision algorithms are employed to detect and locate human faces within images or video frames. Techniques like Haar cascades, Histogram of Oriented Gradients (HOG), and deep learning-based face detectors (such as MTCNN and SSD) are commonly used.
- **Image Preprocessing:** Images containing faces are preprocessed to enhance the features and prepare them for analysis. Preprocessing steps may include resizing, normalization, and color conversion.

Machine Learning and Deep Learning:

- **Feature Extraction:** Deep learning models, particularly Convolutional Neural Networks (CNNs), are used to automatically extract features from the preprocessed face images. These features are crucial for distinguishing between masked and unmasked faces.
- **Model Training:** The extracted features are fed into machine learning or deep learning algorithms for training. During the training process, the model learns to recognize patterns that differentiate between masked and unmasked faces.

- Classification: Once trained, the model can classify new, unseen images (with mask or without mask) based on the learned patterns.

Deployment and Integration:

- Real-time Detection: In real-time applications, such as video surveillance, the trained model is deployed to continuously analyze video frames from cameras. The model detects faces and predicts whether each face is wearing a mask or not in real time.
- User Interface: AI face mask detection systems often include user interfaces to display the detection results. Detected faces are typically marked with bounding boxes, and the output (mask or no mask) is displayed alongside or overlaid on the images or video frames.

Alerts and Notifications:

- Automated Alerts: In certain applications, such as security systems, automated alerts can be generated if individuals without masks are detected. These alerts can trigger notifications to authorities or designated personnel for appropriate actions.
- Data Logging: Data about mask compliance can be logged for further analysis. This data can be used to monitor compliance trends, assess the effectiveness of public health policies, and identify areas where mask-wearing guidelines may need reinforcement.

Challenges and Considerations:

- Accuracy and Robustness: Ensuring the model's accuracy, especially in diverse lighting and environmental conditions, is crucial. Robustness testing is required to handle variations in face poses, mask types, and occlusions.
- Privacy Concerns: Face mask detection systems must be designed with privacy considerations in mind. Anonymizing techniques can be applied to protect individual identities while still ensuring compliance monitoring.
- Ethical Implications: Ethical considerations, such as consent, should be taken into account when deploying face mask detection systems in public spaces. Transparency about data usage and adherence to privacy regulations are essential.
- Real-time Processing: Achieving real-time processing capabilities is vital for applications where immediate responses, such as alerts or notifications, are necessary.

In summary, AI face mask detection technology plays a significant role in promoting public health and safety by automating the monitoring of mask-wearing compliance.

II. REVIEW OF LITERATURE

(1) Effective strategies to restrain the COVID-19 pandemic need high attention to mitigate negatively impacted communal health and the global economy, with the brim-full horizon yet to unfold. In the absence of effective antiviral and limited medical resources, many measures are recommended by WHO to control the infection rate and avoid exhausting the limited medical resources. Wearing a mask is among the non-pharmaceutical intervention measures that can be used to cut the primary source of SARS-CoV2 droplets expelled by an infected individual. Regardless of the discourse on medical resources and diversities in masks, all countries are mandating coverings over the nose and mouth in public. To contribute towards communal health, this paper aims to devise a highly accurate and real-time technique that can efficiently detect non-mask faces in public and thus, enforce wearing masks. The proposed technique is an ensemble of one-stage and two-stage detectors to achieve low inference time and high accuracy. We started with ResNet50 as a baseline and applied the concept of transfer learning to fuse high-level semantic information in multiple feature maps. In addition, we also propose a bounding box transformation to improve localization performance during mask detection. The experiment is conducted with three popular baseline models viz. ResNet50, AlexNet and MobileNet. We explored the possibility of these models to plug in with the proposed model so that highly accurate

results can be achieved in less inference time. It is observed that the proposed technique achieves high accuracy (98.2%) when implemented with ResNet50. Besides, the proposed model generates 11.07% and 6.44% higher precision and recall in mask detection when compared to the recent public baseline model published as the RetinaFaceMask detector. The outstanding performance of the proposed model is highly suitable for video surveillance devices.

(2) The COVID-19 epidemic has swiftly disrupted our day-to-day lives affecting international trade and movements. Wearing a face mask to protect one's face has become the new normal. Soon, many public service providers will expect clients to wear masks appropriately to partake of their services. Therefore, face mask detection has become a critical duty to aid worldwide civilization. This paper provides a simple way to achieve this objective by utilizing some fundamental Machine Learning tools such as TensorFlow, Keras, OpenCV, and Scikit-Learn. The suggested technique successfully recognizes the face in the image or video and then determines whether or not it has a mask on it. As a surveillance job performer, it can also recognize a face together with a mask in motion as well as in a video. The technique attains excellent accuracy. We investigate optimal parameter values for the Convolutional Neural Network model (CNN) to identify the existence of masks accurately without generating over-fitting.

(3) The coronavirus virus which is also called SARS-CoVID-19 is spreading over the world among the people and led to a global pandemic. The spreading of the virus forced the government to pressure their people to follow strict lockdowns, which might cause many problems for every single person in society. The strict rules from the WHO (World Health Organization) state that the only solution for the spread of the virus is wearing a face mask. Thus, to help the government and the public this proposed model ensures that every single person wears a face mask in public places. Few models work depending on artificial intelligence, deep learning, and machine learning. The model uses the Keras, Tensor-flow, and OpenCV methods for execution. This device is developed with two datasets which are namely with and without a facemask. The system is included with a Raspberry- PI camera which captures the live streaming video and covers them into images, which can be used as the inputs and process the data. The system is developed with a toll-way gate which allows only if the person crossing it has a face mask worn on his/her face or it does not allow the person in. It is developed along with an alarm system that beeps with a red light if the person is not wearing a face mask or it glows with a green light. This model ensures to make the environment and the people are safe

(4) The whole world is suffering from a novel coronavirus, which has become an epidemic. According to a World Health Organization report, this is a communicable disease, i.e., it transfers from an infected person to a healthy person. Therefore, wearing a mask is the most important precaution to protect from COVID-19. This paper presented a deep learning-based approach to design a Face Mask Detection framework to predict whether a person is wearing a mask or not. The proposed method uses a Single Shot Multibox detector as a face detector model and a deep Inception V3 architecture (SSDIV3) to extract the pertinent features of images and discriminate them in mask and without mask labels. Optimizing the SSDIV3 approach using different modeling parameters is a genuine contribution of this work. In addition to this, the system is tested and analyzed on VGG16, VGG19, Xception, and Mobilenet V2 models at different modeling parameters. Furthermore, two synthesized novel Face Mask Datasets are introduced containing diversified masks (2d_printed, 3d_printed, handkerchief, transparent, natural-looking mask appearance masks) and unmasked images of humans collected in outdoor and indoor environments such as parks, homes, and laboratories. The experiment outcomes demonstrate that the proposed system has achieved an accuracy of 98% on the synthesized benchmark datasets, which comparatively outperforms other state-of-the-art methods and datasets in a real-time environment.

(5) Since the infectious coronavirus disease (COVID-19) was first reported in Wuhan, it has become a public health problem in China Using Kaggle datasets, the proposed system/model is trained and examined. The system runs in real-time and detects if an individual face has a facemask if not then notify the individual personally through text message. The mask is extracted from real-time faces in public and is fed as an input into a convolutional neural network (CNN).

(6) The COVID-19 pandemic has rapidly increased health crises globally and is affecting our day-to-day lifestyle. A motive for survival recommendations is to wear a safe facemask and stay protected against the transmission of coronavirus. By wearing a facemask, the most effective preventive care must be taken against COVID-19. Monitoring manually if the individuals are wearing facemasks correctly and notifying the victim in public and crowded areas is a difficult task. This paper approaches a simplified way to achieve facemask detection and notify the individual if not wearing a facemask.

(7) In current times, after the rapid expansion and spread of the COVID-19 outbreak globally, people have experienced severe disruption to their daily lives. One idea to manage the outbreak is to force people to wear face masks in public places. Therefore, automated and efficient face detection methods are essential for such enforcement. In this paper, a face mask detection model for static and real-time videos has been presented which classifies the images as “with a mask” and “without a mask”. The model is trained and evaluated using the Kaggle data set. The gathered data set comprises approximately 4,000 pictures and attained a performance accuracy rate of 98%. The proposed model is computationally efficient and precise as compared to DenseNet-121, MobileNet-V2, VGG-19, and Inception-V3. This work can be utilized as a digitized scanning tool in schools, hospitals, banks, airports, and many other public or commercial locations.

(8) The COVID-19 pandemic has rapidly affected our day-to-day life disrupting world trade and movements. Wearing a protective face mask has become a new normal. Soon, many public service providers will ask customers to wear masks correctly to avail of their services. Therefore, face mask detection has become a crucial task to help global society. This paper presents a simplified approach to achieve this purpose using some basic Machine Learning packages like TensorFlow, Keras, OpenCV, and Scikit-Learn. The proposed method detects the face from the image correctly and then identifies if it has a mask on it or not. As a surveillance task performer, it can also detect a face along with a mask in motion. The method attains accuracy up to 95.77% and 94.58% respectively on two different datasets. We explore optimized values of parameters using the Sequential Convolutional Neural Network model to detect the presence of masks correctly without causing over-fitting.

(9) The Covid-19 pandemic has wreaked havoc on people's lives around the world, hurting public health and countries economy. Wearing masks in public areas is now required to limit the risk of contracting the Covid-19 virus. Various public locations, shops, and service providers have now implemented laws requiring visitors to wear masks while on their premises, and if they do not, they will not be able to access or use any of their services. Face mask detection is becoming an increasingly important tool on a larger scale. Using machine learning programs such as Keras, TensorFlow, and OpenCV. The algorithm detects whether or not the person is wearing a mask. To improve the accuracy of the face mask detector, we added photographs of people not wearing masks, as well as people covering their faces with and even around the world. This pandemic is having devastating effects on societies and economies around the world. The increase in the number of COVID-19 tests gives more information about the epidemic's spread, which may lead to the possibility of surrounding it to prevent further infections. However, wearing a face mask that prevents the transmission of droplets in the air maintaining an appropriate physical distance between people, and reducing close contact with each other can still be beneficial in combating this pandemic. Therefore, this research paper focuses on implementing a Face Mask and Social Distancing Detection model as an embedded vision system. The pre-trained models such as the MobileNet, ResNet Classifier, and VGG are used in our context. People violating social distancing or not wearing masks were detected. After implementing and deploying the models, the selected one achieved a confidence score of 100%. This paper also provides a comparative study of different face detection and face mask classification models. One such type of software is Matlab. Hence, this solution tracks the people with or without masks in a real-time scenario and ensures social distancing by generating an alarm if there is a violation in the scene or public places. This can be used with the existing embedded camera infrastructure to enable these analytics which can be applied to various verticals, as well as in an office building or at airport terminals/gates.

objects that aren't masks, such as scarfs, garments, and rags, and covering faces with hands, among other things. This produces more accurate methods for detecting faces without masks, making it more difficult for people to circumvent the face mask detector, as well as the creation of an efficient face mask dataset for future challenges. It can recognize faces wearing masks in real-time.

(10) COVID-19 is a big threat to human mankind. The whole world is now struggling to reduce the spread of the COVID-19 virus. Wearing masks is a good practice that helps to control COVID-19 effectively. From the results of China and South Korea, it is clear wearing a facemask reduces the virus spread. Now they are back to normal life. But ensuring all people wear facemasks is not an easy thing. This paper attempts to develop a simple and effective model for real-time monitoring. The proposed model successfully recognizes if an individual is wearing a face mask or not.

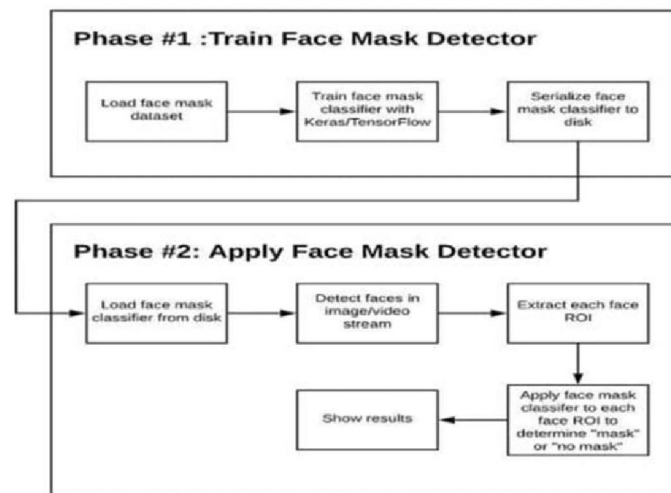
III. DATASET

Step-by-Step Implementation

The Face Mask Detection model is created in four steps

- **Specifying the model:** (layer node, the activation function is applied to those nodes)
- **Compile:** (loss function, Optimizer)
- **Fit:** (make model learn)
- **Predict:** (use the model to predict)

To train a customized face mask detector, we must divide our project into two unique stages, each with its own set of sub-steps (as seen in Figure below):



Two Phases COVID-19 Face Detector

- **Training:** Here we'll focus on loading our face mask detection dataset from disk, training a model (using Keras/Tensor Flow) on this dataset, and then serializing the face mask detector to disk.
- **Deployment:** Once the face mask detector is trained, we can then move on to loading the mask detector, performing face detection, and then classifying each face as with mask or without mask.



IV. PROPOSED METHODOLOGY

1. Data Collection:

To train a face mask detection model, you'll need a dataset of images with and without face masks. You can create your dataset or find existing datasets like the "MaskedFace-Net" dataset or "LFW Masked Face" dataset. Ensure that your dataset is properly labeled to distinguish between masked and unmasked faces.

2. Installing Dependencies:

Make sure you have the required libraries and frameworks installed:

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TensorFlow: A deep learning framework.

Keras: A high-level neural networks API that runs on top of TensorFlow.

OpenCV: An open-source computer vision library for image processing.

3. Preprocessing Data:

Use OpenCV to preprocess the images. You may need to resize, normalize, and augment the data to improve the model's performance.

4. Model Architecture:

Build a Convolutional Neural Network (CNN) using Keras. Your CNN should consist of several convolutional layers, pooling layers, and fully connected layers. This architecture should be designed to learn features that distinguish between masked and unmasked faces.

5. Training:

Split your dataset into training and validation sets. Train your model using the training data, and validate its performance with the validation data. You can use binary classification loss functions like binary cross-entropy.

6. Model Evaluation

After training, evaluate the model on a separate test dataset to assess its accuracy, precision, recall, F1 score, and other relevant metrics.

7. Fine-Tuning and Optimization:

Experiment with different hyperparameters, model architectures, and techniques to improve the model's performance. You may also explore transfer learning by using pre-trained models like MobileNet or VGG.

8. Visualizing Results:

You can use OpenCV to draw bounding boxes around faces and indicate whether a person is wearing a mask or not. This makes the results more interpretable.

9. Deployment:

Deploy your model to the desired platform, whether it's a web application, mobile app, or edge device. TensorFlow provides tools for model deployment, and you may need to consider performance optimization depending on your deployment environment.

10. Monitoring and Maintenance:

Continuously monitor the performance of your face mask detection system and update the model as needed to adapt to changing conditions or new data.

AI face mask detection using TensorFlow, Keras, and OpenCV is a practical application of computer vision and deep learning that can be used in various real-world scenarios, from ensuring public safety during pandemics to improving workplace safety and security.

V. CONCLUSION

In this paper, we have explained the motivation of the work at first. Then, we illustrated the learning and performance tasks of the model. Using basic ML tools and simplified techniques the method has achieved reasonably high accuracy. It can be used for a variety of applications. Wearing a mask may be obligatory shortly, considering the Covid-19 crisis. Many public service providers will ask customers to wear masks correctly to avail of their services. The deployed model will contribute immensely to the public health care system. In the future, it can be extended to detect if a person is wearing the mask properly or not. The model can be further improved to detect if the mask is virus-prone or not i.e. the type of the mask is surgical, N95, or not.

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