



Real-Time Anti Spoofing Face Detection with Mask Using CNN

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Article History	Abstract
Received: 06 June 2023 Revised: 05 Sept 2023 Accepted: 29 Nov 2023	<p>As COVID-19 spread the whole way across the world, a significant number of us got mindful of how significant face covers are. Medical services authorities and nearby foundations from one side of the planet to the other are encouraging individuals to wear masks ,as it is the best way to forestall the transmission of the infection. Masks have without a doubt frustrated the facial-acknowledgment industry; the innovation has likewise adjusted. It might sound odd yet wearing a cover does not really prevent a PC from recognizing somebody. We are intending to prepare our model to recognize whether the pictures are genuine or fake one even though individuals are wearing face cover. In this paper, we intend to make a liveness detector equipped for spotting counterfeit faces. To make a liveness detector, we will prepare a deep learning neural network fit for recognizing genuine versus counterfeit appearances. It deals with two correlative spaces: RGB space and multi-scale Retinex (MSR) space. The RGB space contains the point-by-point facial surfaces, yet it is sensitive to illumination whereas the MSR pictures can adequately catch the high recurrence data, which is discriminative for face recognition.</p> <p>Keywords: Face anti-spoofing, convolutional neural networks, liveness detection, open cv</p>
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1. Introduction

Face recognition applications are becoming more popular than ever before, with applications ranging from face recognition on our smartphones to face recognition for mass surveillance. Face recognition systems are easily fooled by spoofing attacks these days. It is critical to create a face verification structure to effectively protect an individual's privacy. Over the most recent couple of years, Face recognition frameworks have gained popularity in recent years as a result of face-rich features that provide a solid biometric sign to see individuals for a variety of purposes. Face biometrics are gradually being used as a replacement to passwords on mobile phones. Regardless of the remarkable advancements in facial recognition systems, the multifaceted nature of spoofing attacks continues to evolve as more complex counter methodologies are developed, which are efficient, productive, and decreased. Face Spoofing: Face spoofing is an attempt to gain someone else's perks or access rights by imitating someone else's face with a snapshot.

Anti-spoofing: The task of preventing false facial verification by using a snapshot, video, mask, or other alternative replacement for an authorized individual's face is known as facial anti-spoofing. A few instances of assaults:

Print assault: The attacker uses someone's picture in a print assault. The picture is printed or displayed on a high-tech computer. This is the most often seen type of attack, since most peoples have facial pictures available on the internet, and photos could be obtained easily without authorization.

Replay/video assault: A more refined approach to deceiving the framework, which usually includes a circled video of a casualty's face. This strategy ensures immediate and facial changes to appear more 'characteristic' as compared to keeping someone's picture.

3D mask attack: In this type of attack, a mask is used as the parodying apparatus of choice. It is a lot more complicated than just seeing a face frame. Notwithstanding regular facial developments, it

empowers approaches to misdirect some additional layers of insurance like profundity sensors. Photograph assault and video answer trap are the most by and large saw assaults.

Face recognition using OpenCV:

OpenCV is a multi-stage library that can be used to develop ongoing PC vision applications. It primarily focuses on image processing, video capture, and analysis, with features such as face recognition and item location. OpenCV plays a major role in actual event. The OpenCV library has over 2500 updated libraries that can be used to identify and perceive faces, recognize objects, organize human activities in recordings, and monitor camera movements, among other things. To create a face recognition system, we first perform face exploration, then use profound learning to extract face embeddings from each face, then train a face recognition model on the embeddings from each face using deep learning, and then recognize faces in the two pictures and video transfers using OpenCV.

II. Related works

[1] Kunj Bihari Meena proposed a way to distinguish CG photographs from PG pics using a two-flow convolutional neural community (CNN). there are numerous fields inclusive of the movie enterprise, virtual reality, video games where pc-generated (CG) pics are used broadly. CG pictures also can be misused in many approaches. consequently, there may be a want of distinguishing CG pics from real photographic (PG) pics. in the proposed technique, the primary move takes the advantage of the know-how found out by the pre-trained VGG-19 network, and then this know-how is transferred to analyse the distinct capabilities of CG and PG pictures. we advise a 2d stream, that pre-processes the photos the use of 3 excessive-skip filters which intention to help the community to cognizance on noise-based awesome functions of CG and PG photographs. sooner or later, we recommend an ensemble version to merge the outcomes of the proposed two streams.

[2] Surya teja karri proposed 3-D-CNN strategies: MicroExpSTCNN and MicroExpFuseNet, for unconstrained facial miniature demeanour notoriety with the guide of abusing the spatiotemporal records in CNN structure. The MicroExpSTCNN thinks about the entire spatial records, while the MicroExpFuseNet is based absolutely at the 3-d-CNN highlight combination of the eyes and mouth zones. The investigations are executed over CAS(ME) 2 and SMIC microb articulation data sets. The proposed MicroExpSTCNN form outflanks the cutting-edge techniques.

[3] Daqiang Mu proposed a way which utilizes CNN (convolutional neural local area) works as opposed to hand made highlights for face hostile to satirizing. with an end goal to intertwine more prominent discriminative chrominance information, this paper proposes a solitary face hostile to parodying strategy principally dependent on a twofold stream CNN and large demonstrating of highlights from overall face photo and nearby fixes, as well as coordinating the elements of uncommon shade territories, we investigate the discriminative portrayal for face against mocking.

[4] Yongjae Gwak proposed a method for face anti-spoofing based on stereo facial photographs. because the 3-dimensional shape of a live face genuinely yields a structural distinction within the photograph pair taken via a stereo camera, whereas sizable variations do now not occur in faux faces of -dimensional planes, this paper proposes to learn the differences of left-proper image pairs in the latent space of a deep neural network. One critical gain of the proposed method is that the structural distinction is encoded implicitly in a nonlinear way thru the deep architecture without explicitly computing the disparity. The experimental outcomes on a constructed dataset revealed the proposed technique to be effective for numerous spoofing assaults.

[5] Dongjun Yu, proposed a remarkable face against satirizing recognition set of rules utilizing least rectangular weight combination of channel-based element classifiers. To this surrender, we first circuit the tone and surface highlights through data entropy, the spatial and recurrence highlights are then sifted and melded by utilizing SVM-RFE trademark determination approach. likewise, the combination abilities of two convolutional neural organizations are worked through autoencoder. second, for the produced 3 sorts of combination abilities, we receive AdaBoost, SVM and Randomforest to achieve the solid classification, individually. The last objective of the proposed strategies to utilize the most un-rectangular method to change the most reasonable loads of the got 3 assortments of class impacts, through this implies the strong and green face hostile to caricaturing discovery result can be accomplished.

III. COMPARISON AND EXISTING WORKS.

SL. NO	YEAR OF PUBLICATION , AUTHORS AND NAME OF THE PAPER	METHODOLOGY	PERFORMANCE	FUTURE WORK/CHALLENGE
1	2012 J Yang , Sun Z LI Face Liveness Detection Using 3D Structure Recovered From a Single Camera	For a given video or image captured from more than two viewpoints, the facial landmarks are detected and key frames are selected	Support vector machine classifier is trained to distinguish real or fake face	It is difficult to estimate in-depth information and this method is sensitive to noise
2	2012 Andre Arjun LBP-TOP Based Counter Measure Against Face Spoofing Attack	A method to verify faces by Counter measures based on texture and motion. Texture is analyzed using lower binary patterns and motion is analyzed using correlation method	The input video frames are divided into N frames window. Each frame is then analysed based on the motion and micro-texture	User authentication is an important step to protect information

dimensional shape of a live face genuinely yields a structural distinction within the photograph pair taken via a stereo camera, whereas sizable variations do now not occur in faux faces of -dimensional planes, this paper proposes to learn the differences of left-proper image pairs in the latent space of a deep neural network. One critical gain of the proposed method is that the structural distinction is encoded implicitly in a nonlinear way thru the deep architecture without explicitly computing the disparity. The experimental outcomes on a constructed dataset revealed the proposed technique to be effective for numerous spoofing assaults.

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SL. NO	YEAR OF PUBLICATION , AUTHORS AND NAME OF THE PAPER	METHODOLOGY	PERFORMANCE	FUTURE WORK/CHALLENGE
3	2014 Javier Galbally Face Anti-Spoofing Based On General Image Quality Assessment	The input image is first converted to grayscale image. The grayscale image is then filtered using a low-pass Gaussian filter for generating the distorted version of the image.	Comparing the quality between the grayscale and distorted version of the input image by using Image Quality Assessment	Very low degree of complexity
4	2015 D Wen Face Spoof Detection With Image Distortion Algorithm	The features Considered are color diversity, reflection, blurriness and Chromatic moment.	features are trained and Classified using Support Vector Machine (SVM) to identify the face to be either real or spoof face.	It works when the images are captured by web cameras . There is no public-domain face spoof database using mobile phone.

SL. NO	YEAR OF PUBLICATION , AUTHORS AND NAME OF THE PAPER	METHODOLOGY	PERFORMANCE	FUTURE WORK/CHALLENGE
5	2015 K Patel Live Face Video vs Spoof Face Video : Use Of Moire Patterns To Detect Replay Video Attack	In this we analyze the moire pattern aliasing that commonly appears during the recapture of video or photo replays on a screen in different channels (R , G , B and Grayscale) and regions	The moire-patterns can be detected using MLBP and DSIFT features	They capture replay video attacks using either low quality cameras or expensive
6	2016 J Komulainen Face Spoofing Detection Using Colour Texture Analysis	Proposed a methodology to detect the close-up non-real faces by using Histogram of Oriented Gradient (HOG) descriptors.	The alignment of the face is analysed using the upper body detector.	Possibility of printed attacks is more

Sl. No.	YEAR OF PUBLICATION ,AUTHORS AND NAME OF THE PAPER	METHODOLOGY	PERFORMANCE	FUTURE WORK/CHALLENGE
7	2019 I.Ashok Kumar Face Anti Spoofing Using Neural Networks	The convolutional network architecture is constructed to arrest the spoofed faces from accessing in the name of genuine users. Own datasets of real and fake images are created to train the neural network.	The two datasets are trained separately to resolve the absolute outcome.	There's no stable set of features that the convolutional network would see and understand

IV. PROPOSED WORK AND MODULES IDENTIFIED

Spoofing discovery is a classification issue. In profound learning time, a characteristic arrangement of this assignment is to take care of the information RGB pictures to a deliberately planned CNN. In our venture we mean to make a liveness identifier equipped for spotting counterfeit faces and perform against satirizing in face acknowledgment frameworks. To make a liveness indicator, we will prepare a profound neural organization fit for recognizing genuine versus counterfeit appearances. We will attempt to clarify the progression of the arrangement in the following steps:

Step:1: Building the image dataset.

Step:2: Implement LivenessNet.

Step:3: Training the model.

Step:4: Implement it into the real-time.

1) Building the image dataset

In our project we will be training our model using two types of datasets namely: real dataset and fake dataset. For building the real dataset we will record the video using the web camera of the system wherein for the fake dataset we will record the video of a person in our smartphones and then replay it Infront of the web camera. As an input, we'll give our programme the path to these images, and as an output, we'll give it the path to a directory where each of the cropped faces will be saved. We will filter the weak faces and skip the similar adjacent frames. We create blob from an image, define the frame dimensions and extract face ROI coordinates.

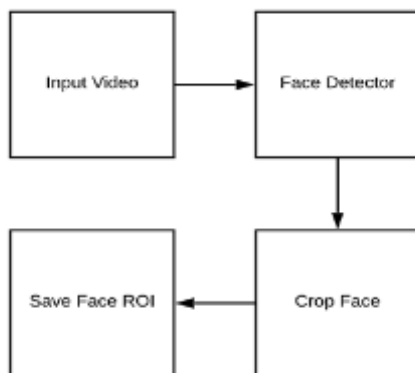


Figure. 1 Diagrammatic flow of dataset collection

2)Implement LivenessNet

Livenessnet is a simple CNN network. In this step we add other layers to our CNN network like :

RELU LAYER- It is an initiation layer.

POOLING LAYER - The pooling layer lessens portion by integrating the yield of neuron packs from one layer into a neuron in the next layer.

FULLY CONNECTED LAYER -completely associated layer interface of each neuron in one layer to each neuron in another layer.

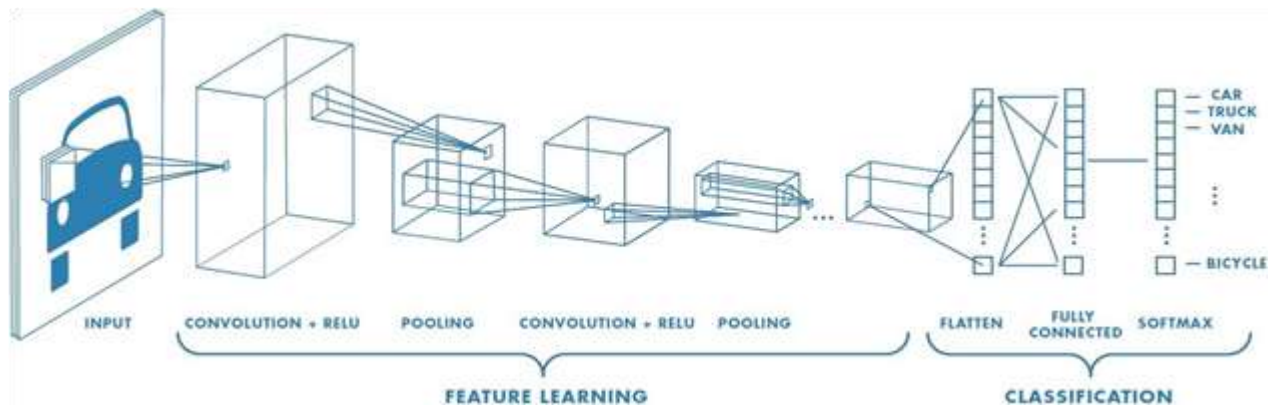


Figure. 2 The various layers of CNN

3) Training the model

We use the datasets of real and fake images and also the liveness Net to train our model. Initially we load the data then resize them to desired pixel values. Each picture will have a label associated with it, which will be stored in the label list. The data must then be partitioned for training and research.

In the end we compile and train our face liveness model.

4) Implementing it into real-time

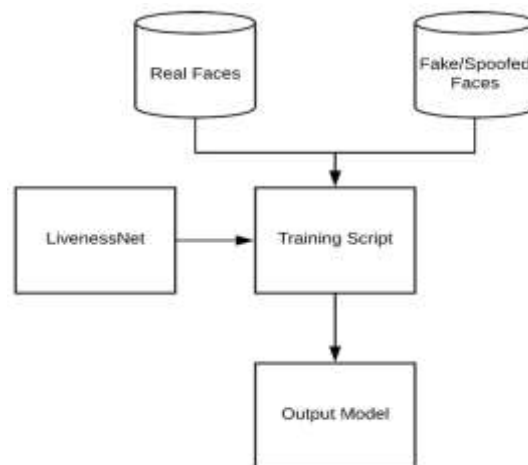


Figure.3 Block diagram

A) Modules Split-up:

The modules identified in the proposed system are as follows:

1. **KERAS**: Keras is an open-source library for neural networks that provide a Python framework.
2. **TENSORFLOW**: TensorFlow is a free and open-source software library for dataflow and distinct programming that can be used to solve a variety of problems.
3. **SKLEARN**: It features support vector machines, random forests, gradient boosting, k-means, and DBSCAN, amongst many other classification, regression, and clustering algorithms, and is designed to work with the Python numerical and science libraries NumPy and SciPy.
4. **MATAPLOTLIB**: Matplotlib is a Python library that allows users to create static, animated, and interactive visualizations.

B) Algorithm:

- 1) Surface analysis, which involves registering Local Binary Patterns (LBPs) over face districts and categorizing the appearances as genuine or satirical using an SVM.
- 2) Recurrence investigation, like as looking at the face's Fourier space.
- 3) Variable centring analysis, such as investigating the range of pixel values between two sequential edges.
- 4) Calculations based on generalizations, such as eye growth, lip development, and flicker position. These calculations help to monitor the client's eye development and squints to ensure that the customer

is not keeping an image of anybody else (since a photograph does not really flicker or move its lips) [17-21].

SYSTEM REQUIREMENTS

Hardware Requirements: Windows 10(i3 processor),

8 GB RAM

Software Requirements: Python

PyCharm

Anaconda 3

Modules: Open CV (cv2) and NumPy

3. Results and Discussion

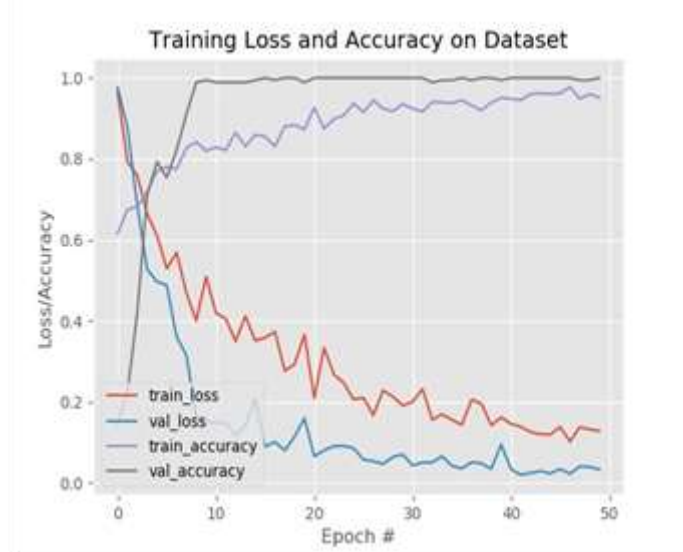


Figure.4 Loss graph

The above graph shows the accuracy of training in four ways. Screen shots of data sets: We have two datasets namely real images and fake images.

Real: Real video is given as input and the program is written to prepare the data set like this where the images are cropped and stored in the same size.

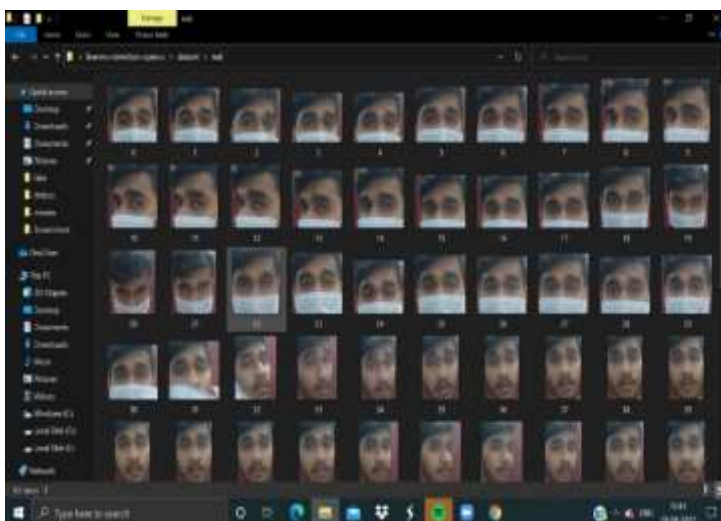


Figure.5 Real images dataset

Fake: Fake video is given as input and the program is written to prepare the data set like this where the images are cropped and stored in the same size.

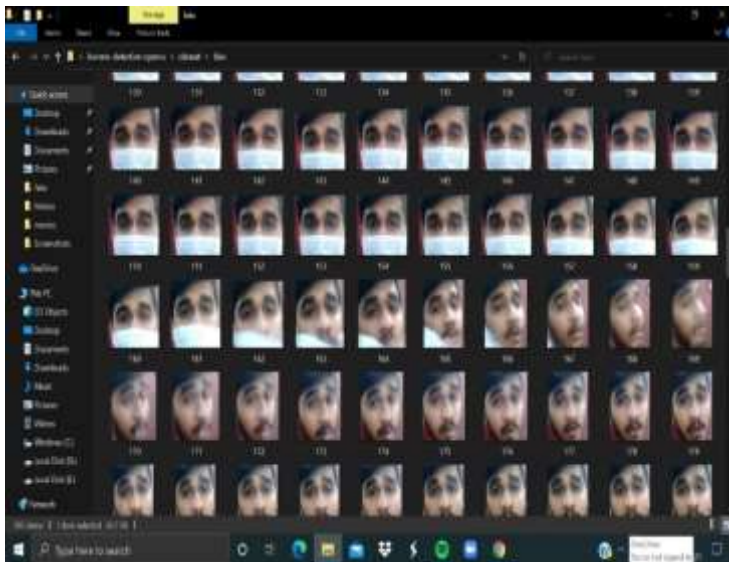


Figure.6 Fake images dataset

Analysis:

(To analyse whether the images are real or spoof)

For real image:

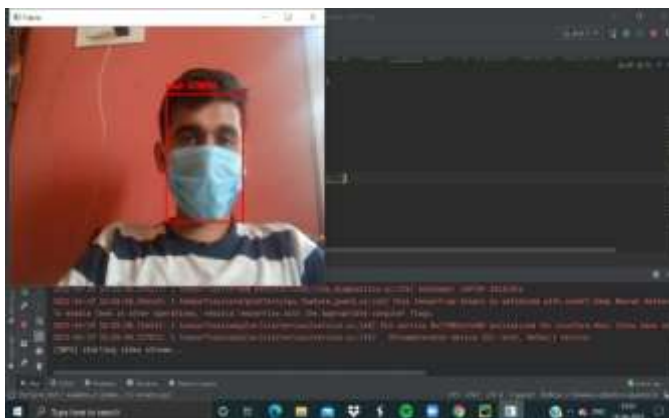
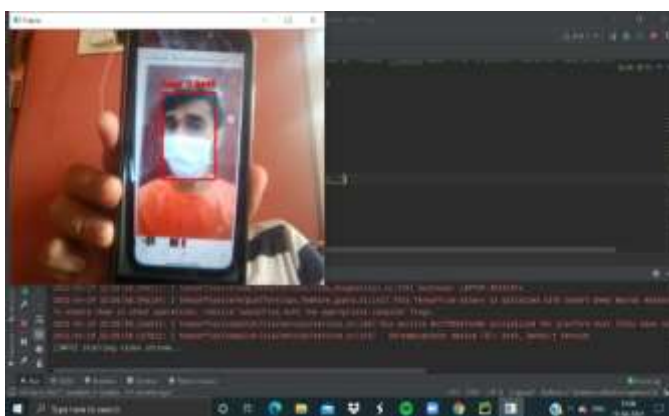


Figure.7 Detection of real image

As you can see in the figure a real face is detected.

For spoof image:



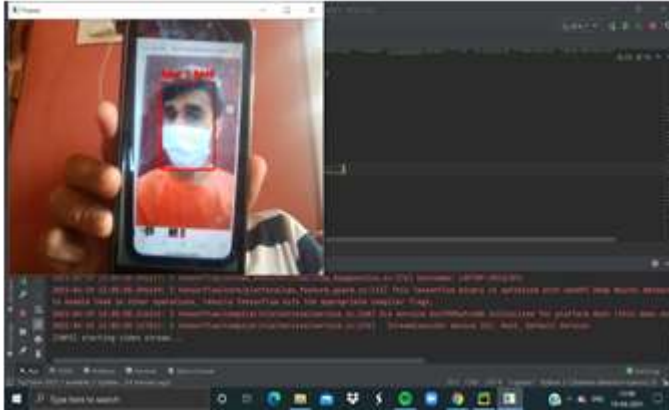


Figure.8 Detection of fake image

If the person tries to spoof using photo masking or video masking techniques then the result will be shown as fake, as you see in the figure 6.

4. Conclusion

To adequately defend an individual's privacy, face recognition systems must always be made more accessible, and we must be able to identify fake faces. We used OpenCV to detect liveness in order to accomplish this. We developed a liveness detector for face recognition systems that can detect fake faces and perform anti-face spoofing. We used OpenCV, TensorFlow, and Keras libraries. We have repeatedly prepared the model with data collection to framework using Keras and sklearn modules.

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