

Attendance Prediction Based on Weather Condition

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Abstract:

Since the early beginnings of education systems, attendance has always played a crucial role in student success, as well as in the overall interest of the matter. The most productive way of increasing the student attendance rate is to understand why it decreases, try to predict when it is going to happen, and act on causing factors in order to prevent it. Many benefits of predicted and increased attendance rate can be achieved, including better lecture organization (i.e., lecture time and duration, lecture class choice, etc.). This project describes the steps in the extraction of knowledge from the university's student database and making a model that predicts whether the student will attend the class or not based on weather. In this project attendance patterns are reflected using a Harr cascade algorithm.

Keywords—Weather Forecast, Attendance Prediction, Machine Learning, Face Detection, Convolutional Neural Network, Tensor Flow.

I. INTRODUCTION

Weather forecasting is the application of science and technology to predict the state of the atmosphere for a given location and current location based on the data of estimated rainfall. Ancient weather forecasting methods usually relied on observed patterns of events, also termed pattern recognition. It might be observed that if the sunset was particularly red, the following day often brought fair weather. This includes temperature, rain, cloudiness, wind speed, and humidity. However, not all of these predictions prove reliable. Weather warnings are a special kind of short-range forecast carried out for the protection of human life. Weather warnings are issued by governments throughout the world for all kinds of threatening weather events including tropical storms and tropical cyclones depending upon the location distinguishing a specific group

of entities i.e., Face. It has numerous applications, such as education, surveillance, and so on. This paper presents a simplified approach to serve the above purpose using the basic Machine Learning (ML) packages such as TensorFlow, Keras, OpenCV and Scikit-Learn

II. RELATED WORK

In face detection method, a face is detected from an image that has several attributes in it. According to research into face detection requires expression recognition, face tracking, and pose estimation. Given a solitary image, the challenge is to identify the face from the picture. Face detection is a difficult task because the faces change in size, shape, color, etc. and they are not immutable. It becomes a laborious job for opaque image by some other thing not confronting camera, and so forth.

Authors in think occlusive face detection comes with two major challenges:

1) unavailability of sizably voluminous datasets containing both masked and unmasked faces

2) exclusion of facial expression in the covered area. Utilizing the locally linear embedding (LLE) algorithm and the dictionaries trained on an immensely colossal pool of masked faces, synthesized mundane faces, several islaid expressions can be recuperated, and the ascendancy of facial cues can be mitigated to great extent. According to the work reported in, convolutional neural network (CNNs) in computer vision comes with a strict constraint regarding the size of the input image. The prevalent practice reconfigures the images before fitting them into the network to surmount the inhibition.

Here the main challenge of the task is to detect the face from the image correctly and then identify if it has a mask on it or not. To perform surveillance tasks, the proposed method should also detect a face along with a mask in motion.

III. DATASET

Two datasets have been used for experimenting the current method. Dataset 1 consists of trained images of students. Fig. 1 shows the samples of students trained pictures.



Fig. 1: Samples from Dataset 1

Dataset 2 consists of images of students with names and id for better attendance. Fig. 2 shows trained and tested faces of students.

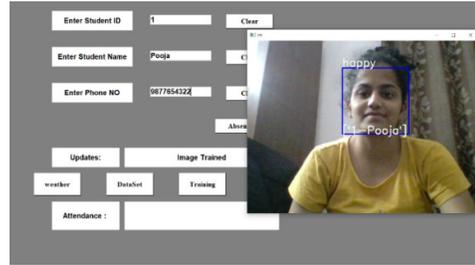


Fig. 2: Samples for Dataset 2 of student's face recognition

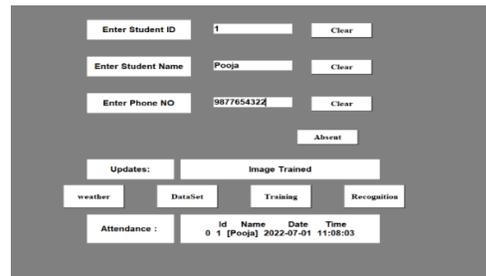


Fig 3: Sample for student Recording Attendance

IV. INCORPORATED PACKAGES

A. TensorFlow

TensorFlow, an interface for expressing machine learning algorithms, is utilized for implementing ML systems into fabrication over a bunch of areas of computer science, including sentiment analysis, voice recognition, geographic information extraction, computer vision, text summarization, information retrieval, computational drug discovery and flaw detection to pursue research. In the proposed model, the whole Sequential CNN architecture (consists of several layers) uses TensorFlow at backend. It is also used to reshape the data (image) in the data processing.

B. Keras

Keras gives fundamental reflections and building units for creation and transportation of ML arrangements with high iteration velocity. It takes full advantage of these

alability and cross-platform capabilities of TensorFlow. The core data structures of Keras are layers and models. All the layers used in the CNN model are implemented using Keras. Along with the conversion of the class vector to the binary class matrix in data processing, it helps to compile the overall model.

C. OpenCV

OpenCV (Open-Source Computer Vision Library), an open-source computer vision and ML software library, is utilized to differentiate and recognize faces, recognize objects, group movements in recordings, trace progressive modules, follow eye gesture, track camera actions, expel red eyes from pictures taken utilizing flash, find comparative pictures from an image database, perceive landscape, and set up markers to overlay it with increased reality and so forth. The proposed method makes use of these features of OpenCV in resizing and color conversion of data images.

V. THE PROPOSED METHOD

The proposed method consists of a cascade classifier and a pre-trained CNN which contains two 2D convolution layers connected to layers of dense neurons. The algorithm for face mask detection is as follows:

A. Data Processing

Data preprocessing involves conversion of data from a given format to much more user-friendly, desired, and meaningful format. It can be in any form like tables, images, videos, graphs, etc. This organized information fits within an information model or composition and captures relationship between different entities. The proposed method deals with image and video data using NumPy and OpenCV.

B. Data Visualization

Data visualization is the process of transforming abstract data to meaningful representations using

knowledge communication and insight discovery through encodings. It is helpful to study a particular pattern in the student image dataset. Data visualization provides a good, organized pictorial representation of the data which makes it easier to understand, observe, analyse.

C. Conversion of RGB image to gray image

Modern descriptor-based image recognition systems regularly work on grayscale images, without elaborating the method used to convert from color-to-grayscale. This is because the color-to-grayscale method is of little consequence when using robust descriptors. Introducing non-essential information could increase the size of training data required to achieve good performance. As grayscale rationalizes the algorithm and diminishes the computational requisites, it is utilized for extracting descriptors instead of working on color images instantaneously.

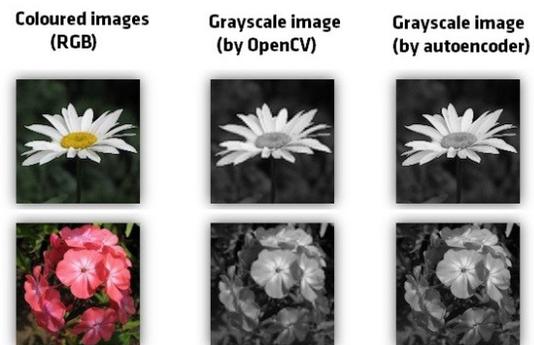


Fig.3. Conversion of a RGB image to a Gray Scale images

We use the function `cv2.cvtColor(input image, flag)` for changing the color space. Here flag determines the type of conversion. In this case, the flag `cv2.COLOR_BGR2GRAY` is used for gray conversion.

D. Image Reshaping

The input during relevation of an image is a three-dimensional tensor, where each channel has a prominent unique pixel. All the images must have identically tant amount size corresponding to 3D feature tensor. How-ever, neither images are

customarily coextensive nor their corresponding feature tensors. Most CNNs can only accept fine-tuned images. This engenders several problems throughout data collection and implementation of model. However, reconfiguring the input images before augmenting them into the network can help to surmount this constraint.

The images are normalized to converge the pixel range between 0 and 1. Then they are converted to 4-dimensional arrays using $\text{data} = \text{np.reshape}(\text{data}, (\text{data.shape}[0], \text{img size}, \text{img size}, 1))$ where 1 indicates the Grayscale image. As the final layer of the neural network has 2 outputs – with mask and without mask i.e., it has categorical representation, the data is converted to categorical labels.

E. Training of Model

Building the model using CNN architecture: Deep learning is a very significant subset of machine learning because of its high performance across various domains. Convolutional Neural Network (CNN), is a powerful image processing deep learning type often used in computer vision that comprises an image and video recognition along with a recommender system and natural language processing (NLP).

The CNNs have several different filters/kernels consisting of trainable parameters which can convolve on a given image spatially to detect features like edges and shapes. These high number of filters essentially learn to capture spatial features from the image based on the learned weights through back propagation and stacked layers of filters can be used to detect complex spatial shapes from the spatial features at every subsequent level.

Fig.4.Face Recognition using CNN

CNN architecture for face recognition system is proposed including the process of collecting face data of students. Experimentally it is shown that the proposed CNN architecture provides 99% accuracy. Further, the proposed CNN framework is used to develop a “Smart Attendance Management System (SAMS)”, which is a web-based application, to provide attendance of students using face recognition, in realtime. The proposed application is easy to deploy and maintain.

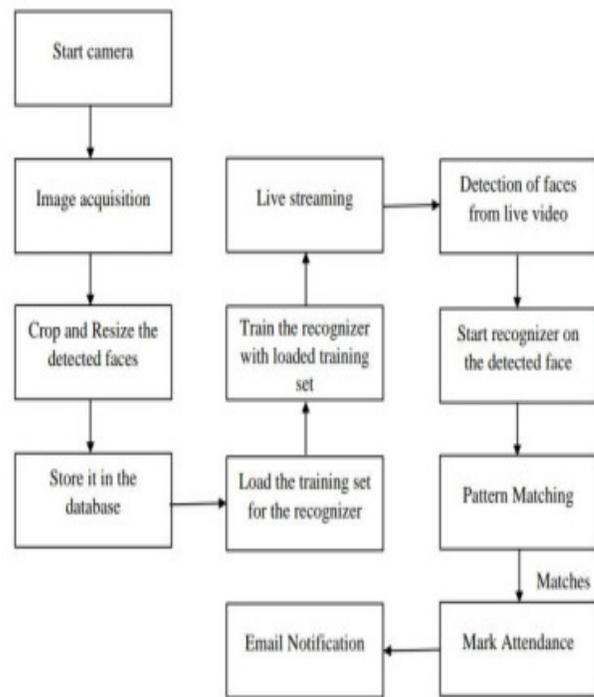
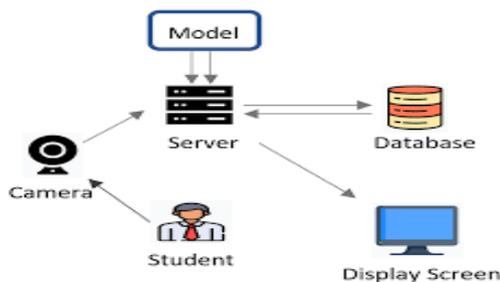


Fig.5. Overview of the Model



VI. RESULT AND ANALYSIS

The users can interact with the system using a GUI. Here users will be mainly provided with three different options such as, student registration, faculty registration, and mark attendance. The students are supposed to enter all the required details in the student registration form. After clicking on register button, the web

cam starts automatically and pops up and starts detecting the faces in the frame. Then it automatically starts clicking photos until 60 samples are collected or CRTL+Q is pressed. These images then will be pre-processed and stored in training images folder.

Attendance prediction based on the weather forecast is the one in which the organization can analyze or predict whether the student will attend the class based on present weather condition.

	A	B	C	D	E	F	G	H	I	J
1	Id	name	phone							
2	1	Pooja	9.88E+09							
3	2	nisha	9.89E+08							
4	3	navya	9.76E+10							
5										
6										
7										
8										
9										
10										
11										

Fig 6:Attendance sheet

VII. CONCLUSIONS

In this paper, system aims to build an effective class attendance system using face recognition techniques. The proposed system will be able to mark the attendance via face Id. It will detect faces via webcam and then recognize the faces. After recognition, it will mark the attendance of the recognized student and update the attendance record.

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