

Automatic attendance-System Using Face Recognition System

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Abstract

Despite huge latest advances in the subject of face reputation, enforcing face verification and popularity adroitly at scale offers extreme challenges to modern-day processes. here we gift a machine, referred to as FaceNet that without delay learns a aligning from face photos to a firm Euclidean space in which distances precisely correlate to a degree of face affinity. as soon as this area has been composed, responsibilities such as face popularity, verification and clustering may be effortlessly carried out the usage of standard techniques with FaceNet embedding's as feature trajectory. Our approach uses a deep conventional community educated to at once enhance the embedding itself, in place of an transitional hindrance layer as in preceding wide mastering methods.

1. INTRODUCTION

FACE popularity is an critical studies problem spanning several fields and disciplines. This due to the fact face popularity, in auxiliary to having diverse realistic applications which include affinity card identity, get entry to curb, Mug photographs searching, protection monitoring, and surveillance device, is a rudimentary human reaction this is imperative for fruitful communications and intercourse amongst human beings. A authorized technique of catalogue faces was initially proposed in June 21, 1888.. the author submitted cluster facial portrait as arc, locating their criterion, and then classifying other portrait by means of their variances from the scale. This category is multi-modal, culminate in a track of self-reliant measures that might be in comparison with other trajectory in a storage. Development has superior to the factor that face popularity systems are being validated in physical-global framework. The speedy improvement of face reputation is due to a mixture of things: lively improvement of design, the opportunity of a huge storage of facial pix, and a way for grading the overall achievement of face recognition designs.

in the literature's, face reputation obstacle can be devised as: given static or recorded pics of a arena, pick out or affirm one or greater individuals inside the arena by means of evaluating with faces saved in a storage. Whilst analyzing man or woman authentication to stand recognition, there are disparate factors which vary. First, a user – an authorized person of a intimate identity machine – is thought to be collegial and makes an identification suit. Compatatibly because of this it isn't crucial to consider the entire set of storage images as a way to authenticate a claim. An incoming photograph is as a consequence in comparison to a short number of model pics of the individual whose identification is said and no longer, as inside the popularity state of affairs, with every photo in a probably huge storage. secondly, an automated authentication system have to function in near-actual time to be applicable to customers. eventually, in popularity attempts, simplest pics of humans from the schooling storage are provided to the gadget, whereas the case of an impostor (most in all likelihood a formerly unseen man or woman) is of maximum importance for authentication. Face popularity is a biometric technique that students computerized strategies to confirm or understand the identification of a dwelling character based on their corporal characteristics. In fashionable, a biometric identity device makes use of either corporal characteristics or behaviour patterns to identify a person. due to human ingrained provocativeness of their eyes, a few people are afraid to apply eye identity systems. Face popularity has the gain of being a apathetic, non nosy system to check non-public status in a “natural” and affable way.

2. LITERATURE SURVEY

On this bankruptcy offers an outline on the foremost human face reputation skills that cover mainly to frontispiece faces, benefits and drawbacks of each technique also are accustomed.

The tactics are consider in circumstances of the facial portrayal they used. on this chapter offers an audit on the most important human face reputation procedures that cover typically to frontispiece faces, benefits and cons of each design are also accustomed. The strategies taken into consideration are eigenfaces , neural networks, dynamic link architecture, hidden Markov version, geometrical characteristic matching, and template matching. The methods are analyzed in terms of the facial representations they used.

2.1 LITERATURE REVIEW:

A. Eigenfaces

Eigen face is special of the maximum very well explored tactics to face popularity. it is also called Karhunen- Loève enlargement, eigen picture, eigen vector, and predominant issue. Allusion Low-Dimensional technique for the characterization of individual frontal, application of the Karhunen-Loève system for the characterization of individual frontal used principal thing evaluation to efficaciously constitute snap shots of frontal. They contend that any frontal picture can be generally reassembled with the aid of a short series of hefts for each frontal and a well-known face image (eigen photograph). The hefts depicting each frontal are acquired with the aid of projecting the face pic onto the eigen image. Eigenfaces for reputation used eigenfaces, which become encouraged by means of the method of Kirby and Sirovich, for face detection and identity.

B. Neural Networks

Counterfeit neural systems (ANNs) are programming usage of the neuronal structure of our cerebrums. We don't have to discuss the unpredictable science of our cerebrum structures, yet get the job done to say, the mind contains neurons which are somewhat resembles to natural switches. These can modify their yield state unpredictable upon the calibre of their electrical or substance input. The neural system in an individual's cerebrum is a massively interconnected system of neurons, where the yield of some random neuron might be the contribution to a huge number of different neurons. Learning happens by more than once initiating certain neural associations over others, and this strengthens those associations. This makes them bound to deliver an ideal result given a predetermined information. This learning involve input – when the ideal result happens, the neural collaboration generating that result become strengthened. Artificial neural systems endeavor to streamline and imitate this cerebrum conduct. They can be composed in a constrain or unsupervised way. In a directed ANN, the system is prepared by giving coordinated info and yield information tests, with the aim of getting the ANN to give an ideal yield to a given information. A model is an email spam channel – the information preparing information could be the include of different words in the body of the email, and the yield preparing information would be a characterization of whether the email was genuinely spam or not. On the off chance that numerous instances of messages are gone through the neural system this enables the system to realize what input information makes it likely that an email is spam or not. The organic neuron is reproduced in an ANN by an enactment work. In characterization assignments (for example recognizing spam messages) this actuation work must have a "switch on" trademark – at the end of the day, when the info is more prominent than a specific esteem, the yield should change state for example from 0 to 1, from - 1 to 1 or from 0 to >0.ple.

2.2 Working

It Works by using the two nueral networks where each neural crop divide and align and the image is identified by the amount of shading percentage that ranges from 0 to 1. Firstly, we need to create two neurals and 128 dimensional text file where we can store all the cordinates of the face with points ranges from negative to positive and we need to create the data base where the attendance of the

student is stored with the respect to the details which are predefined attributes where the student number will be primary key.

```
CREATE TABLE IF NOT EXISTS `studentattendencedetails` (  
  `Rno` int(11) NOT NULL AUTO_INCREMENT,  
  `Sno` int(11) NOT NULL,  
  `Sname` varchar(500) NOT NULL,  
  `Adate` date NOT NULL,  
  `RecordDate` date NOT NULL,  
  PRIMARY KEY (`Rno`)  
)
```

We need to activate the anaconda environment which will connect to the server where we can access all the python libraries. so the camera is used to record the data of individual persons with some points at some Euclidian distance. So while recording the data the respective person's identity is entered so it is easy to connect the storage data with mysql database. The recorded data is stored in 128d.text with their coordinates and with their respective names. All the recorded data is stored in only one text file which can be differentiated. While we start recognising the faces the input image is verified and clustered by dividing the image into many minute divisions. So it can be mapped with already clustered divisions and recognize the individual faces by mapping the clusters, when the face is recognized it will show the percentage of recognition and the attendance is updated with respect to the attribute called name for every time interval till the session expires. And the data is inserted and updated in mysql database, with the data and time of the recognition.

```
CREATE TABLE IF NOT EXISTS `studentdetails` (  
  `Sno` int(11) NOT NULL AUTO_INCREMENT,  
  `Sname` varchar(500) NOT NULL,  
  `course` varchar(100) NOT NULL,  
  `Semister` int(11) NOT NULL,  
  `AttendedDay` int(11) NOT NULL,  
  `Reviews` varchar(1000) NOT NULL,  
  PRIMARY KEY (`Sno`)  
)
```

3. IMPLEMENTATION

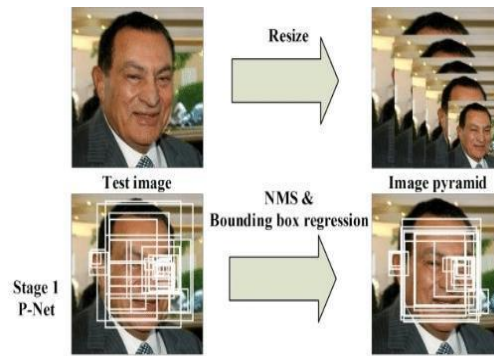
3.1 PROBLEM DEFINITION:

by the usage of this software the time constraint taken by using the faculty to take all the scholar attendance can be reduced and it makes more smooth that the faculty need now not waste time within the instructions, if the group had data of all of the students then they can effortlessly get their statistics.

3.2 MODULES DESCRIPTION:

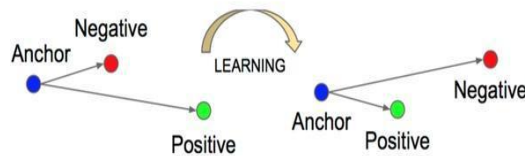
MTCNN Algorithm

The MTCNN set of rules works in three steps and use one neural network for each. the first element is a suggestion community. it will expect capacity face positions and their bounding bins like an interest network in quicker R-CNN. The end result of this step is a large number of face detections and lots of false detections. the second one component uses pics and outputs of the primary prediction. It makes a refinement of the result to do away with maximum of false detections and combination bounding bins. The ultimate part improves even greater the forecasts and adds facialpieces landmarks predictions (inside the unique MTCNN implementation).



Facenet

Face net is a one-shot version, that sequentially learns a matching from face recordings to a condense Euclidean distance in which distances at once correlate to a degree of face resemblance. once this capacity has been generated, tasks which includes face popularity, verification and classifying may be easily applied the use of widespread strategies with FaceNet embeddings as function trajectory(from the original paper). To train, they used three of approximately lineup matching / non-matching facepatches. A triplet is not anything but a collection one anchor photo, one matching photo to the anchor photo and one non-matching picture to the person photo. So the triplet loss minimises the gap among an anchor and a superb, each of that have the equal identity, and maximises the space between the anchor and a con of a distinctive identity.



we can use a pre-skilled version for our use case. here are the steps.

Collect the Picture.

1. Record the images by using mtcnn and save them in the 128d.txt file where all the coordinates are saved.
2. Mention the names of the recorded images while doing the input process.
3. connect the input with database where the persons name is matched with names of the recorded object.

Here we have got the entity of 128-dimensional insertings with corresponding person attribute value. while ever an consumer faces your detection camera, the picture being captured might be ran through the pre-skilled network to create the 128-dimensional embedding for you to then be compared to the saved embeddings the usage of euclidean(L2) distance.

3.3 SYSTEM ARCHITECTURE

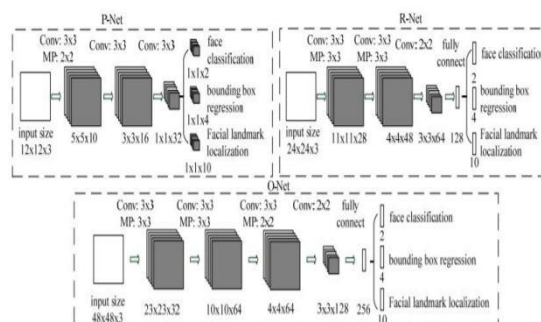
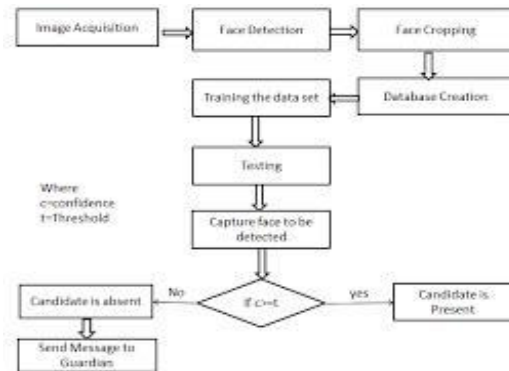


Fig. 2. The architectures of P-Net, R-Net, and O-Net, where "MP" means max pooling and "Conv" means convolution. The step size in convolution and pooling is 1 and 2, respectively.

Flow Charts for Modules:



4.UML DIAGRAMS

4.1 UNIFIED MODELLING LANGUAGE:

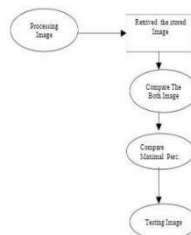
The Unified Modeling Language lets in the software program engineer to explicit a study model using the modeling codes this is administered through a set of linguistic semantic and logical policies. A UML machine is described using five one of a kind aspects that represent the system from extraordinarily unique context. every view is described with the aid of a set of diagram, that is as given.

DATA FLOW DIAGRAM

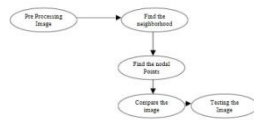
LEVEL 0



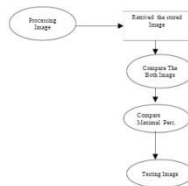
LEVEL 1:PRE PROCESSING



PROCESSING

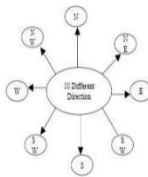


RECOGNITION



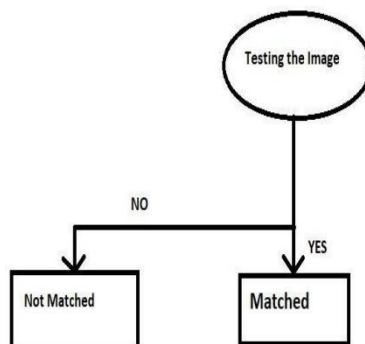
LEVEL 2

10 DIFFERENT DIRECTION



LEVEL 3

TESTING THE IMAGE

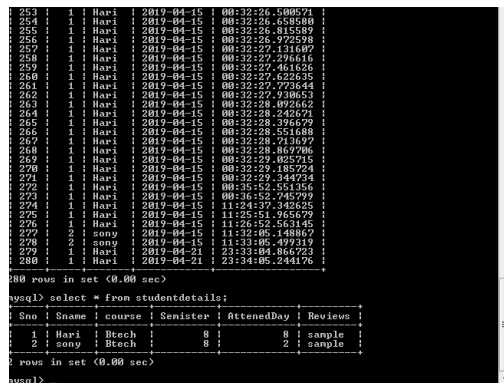


5.RESULT



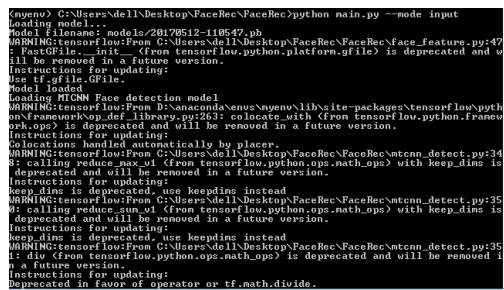
```
C:\Users\del1\Desktop\FaceRec\FaceRec>python main.py
Loading model...
Model filename: models\20170512-110547.pb
WARNING:tensorflow:From C:\Users\del1\Desktop\FaceRec\FaceRec\Face_Feature.py:47:
FactFile.__init__ (from tensorflow.python.platform.gfile) is deprecated and will
be removed in a future version.
Instructions for updating:
Use tf.gfile.GFile.
Model loaded
Loading MTCNN Face detection model
WARNING:tensorflow:From D:\anaconda\envs\myenv\lib\site-packages\tensorflow\pyth
on\framework\op_def_library.py:263: colocate_with (from tensorflow.python.frame
work.ops) is deprecated and will be removed in a future version.
Instructions for updating:
Colocations handled automatically by placer.
WARNING:tensorflow:From C:\Users\del1\Desktop\FaceRec\FaceRec\attnn_detect.py:24:
tf.calling reduce_max_vl (from tensorflow.python.ops.math_ops) with keep_dims is
deprecated and will be removed in a future version.
Instructions for updating:
keep_dims is deprecated, use keepdims instead
WARNING:tensorflow:From C:\Users\del1\Desktop\FaceRec\FaceRec\attnn_detect.py:25:
tf.calling reduce_sum_vl (from tensorflow.python.ops.math_ops) with keep_dims is
deprecated and will be removed in a future version.
Instructions for updating:
keep_dims is deprecated, use keepdims instead
WARNING:tensorflow:From C:\Users\del1\Desktop\FaceRec\FaceRec\attnn_detect.py:25:
tf.div (from tensorflow.python.ops.math_ops) is deprecated and will be removed i
n a future version.
Instructions for updating:
Use tf.divide instead.
WARNING:tensorflow:From C:\Users\del1\Desktop\FaceRec\FaceRec\attnn_detect.py:25:
tf.div (from tensorflow.python.ops.math_ops) is deprecated and will be removed i
n a future version.
Instructions for updating:
Use tf.divide instead.
MTCNN Model loaded
[INFO] camera sensor warning up...
[WARNING] terminating async callback
name found -> hari <class 'str'>
[('Hari', 'Btech', 8, 7, 'sample')]
insert into studentattendancedetails(Sno, Sname, Sdate, ATime) values('Hari', '201
9-04-21', '23:33:04.866723')
inserted
name found -> hari <class 'str'>
[('Hari', 'Btech', 8, 7, 'sample')]
insert into studentattendancedetails(Sno, Sname, Sdate, ATime) values('Hari', '201
9-04-21', '23:34:05.244176')
inserted
C:\Users\del1\Desktop\FaceRec\FaceRec>
```

Fig: Connecting to the anaconda environment and recognition of face and updating the attendance



```
253 | 1 | Hari | 2019-04-15 | 00:32:26.599571 |
254 | 1 | Hari | 2019-04-15 | 00:32:26.558580 |
255 | 1 | Hari | 2019-04-15 | 00:32:26.815389 |
256 | 1 | Hari | 2019-04-15 | 00:32:26.972598 |
257 | 1 | Hari | 2019-04-15 | 00:32:27.115407 |
258 | 1 | Hari | 2019-04-15 | 00:32:27.296416 |
259 | 1 | Hari | 2019-04-15 | 00:32:27.461626 |
260 | 1 | Hari | 2019-04-15 | 00:32:27.626835 |
261 | 1 | Hari | 2019-04-15 | 00:32:27.792044 |
262 | 1 | Hari | 2019-04-15 | 00:32:27.957253 |
263 | 1 | Hari | 2019-04-15 | 00:32:28.092662 |
264 | 1 | Hari | 2019-04-15 | 00:32:28.242671 |
265 | 1 | Hari | 2019-04-15 | 00:32:28.392680 |
266 | 1 | Hari | 2019-04-15 | 00:32:28.551688 |
267 | 1 | Hari | 2019-04-15 | 00:32:28.701697 |
268 | 1 | Hari | 2019-04-15 | 00:32:28.867906 |
269 | 1 | Hari | 2019-04-15 | 00:32:29.023115 |
270 | 1 | Hari | 2019-04-15 | 00:32:29.185724 |
271 | 1 | Hari | 2019-04-15 | 00:32:29.344734 |
272 | 1 | Hari | 2019-04-15 | 00:32:29.503743 |
273 | 1 | Hari | 2019-04-15 | 00:36:52.745799 |
274 | 1 | Hari | 2019-04-15 | 11:24:57.342643 |
275 | 1 | Hari | 2019-04-15 | 11:25:51.965679 |
276 | 1 | Hari | 2019-04-15 | 11:26:52.583145 |
277 | 2 | sony | 2019-04-15 | 11:32:05.148869 |
278 | 2 | sony | 2019-04-15 | 11:33:05.499319 |
279 | 1 | Hari | 2019-04-21 | 23:33:04.866723 |
280 | 1 | Hari | 2019-04-21 | 23:34:05.244176 |
280 rows in set (0.00 sec)
mysql> select * from studentdetails;
+----+-----+-----+-----+-----+-----+
| Sno | Sname | course | Semester | AttendedDay | Reviews |
+----+-----+-----+-----+-----+-----+
| 1 | Hari | Btech | 8 | 8 | sample |
| 2 | sony | Btech | 8 | 2 | sample |
+----+-----+-----+-----+-----+-----+
2 rows in set (0.00 sec)
mysql>
```

Fig: The percentage the faces matched with the images already stored



```
C:\Users\del1\Desktop\FaceRec\FaceRec>python main.py --mode input
Loading model...
Model filename: models\20170512-110547.pb
WARNING:tensorflow:From C:\Users\del1\Desktop\FaceRec\FaceRec\Face_Feature.py:47:
FactFile.__init__ (from tensorflow.python.platform.gfile) is deprecated and will
be removed in a future version.
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Model loaded
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on\framework\op_def_library.py:263: colocate_with (from tensorflow.python.frame
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Colocations handled automatically by placer.
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n a future version.
Instructions for updating:
Use tf.divide instead.
WARNING:tensorflow:From C:\Users\del1\Desktop\FaceRec\FaceRec\attnn_detect.py:25:
tf.div (from tensorflow.python.ops.math_ops) is deprecated and will be removed i
n a future version.
Instructions for updating:
Use tf.divide instead.
```

Fig: Recording the face to store as input

7. LIMITATIONS AND FUTURE ENHANCEMENT

Multi-spectral decipher is the manner of synchronal procurement of a hard and fast of snap shots. The each picture in the multi-spectral decipher resembles to a specific band in the electromagnetic

spectrum. The plain instance of multispectral photo is coloration image inside the visible spectrum having RGB sensations and also can be determined by using human eyes. The hyperspectral pictures consist of more levels in a selected sub-band in comparison to the multispectral photos. Hyperspectral Imaging system (HIS) presents beneficial discriminants for character FR that can't be carried out with the aid of any present imaging gadget. A hyperspectral FR procedure examine spectral curves for the different facial tissue in NIR range of electromagnetic spectrum²⁵. A hyperspectral imaging database of 31 stages of 200 subjects become developed the usage of CCD (rate Coupled gadgets) digital camera, having Liquid Crystal Tunable filter (LCTF). Spectral curves produced in this way display reflectance of subsurface tissue structure which varies for every character, but stable for an extended period. The spectral traits of person tissue aren't stricken by face course and look which lets in to apply the hyperspectral discriminants for FR over a huge style of poses and look

6. CONCLUSION

five(5) specific processes in face detection and popularity were reviewed, particularly, PCA, LDA, skin shade, Wavelet and synthetic Neural community. There are 4 parameters that are taken into consideration on this evaluation, that are size and sorts of database, illumination tolerance, facial expressions versions and pose variations. From this independent assessment, please note that the outcomes ordinary and variant as they correspond to one of a kind experiments or research done by way of previous researchers. as a result, no precise justification can be made as a end on which algorithm is the best for precise duties or mission which include numerous databases, numerous poses, illumination tolerance and facial expressions versions. The performance of the algorithms relies upon on severa elements to be taken under consideration. in place of the usage of those algorithms entirely, they may be stepped forward or more desirable to end up a new technique or hybrid approach that yields a better overall performance.

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