ONLINE EXAMINATION WITH FACE RECOGNITION

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ABSTRACT - In this paper, we present a benchmark of the Local Binary Patterns Histogram (LBPH) algorithm together with Open CV and Python. We wanted to indicate that these techniques are often used for a real-time application. The most a part of this work will describe the architecture similarly because the implementation of the detection and recognition application.

Keywords - Real-time Systems, Face Detection, Face Recognition, Benchmark.

I. INTRODUCTION

Face detection and face recognition are often employed in various real-world application fields. This might be face tracking during a live broadcast of sport event or person identification at the airport border control. Another field of application can be action publicly places using the image data of surveillance cameras. Both examples need a quick detection and recognition algorithm because the results of detecting and recognizing a face are important for further processing.

A. Face Detection

The primary interface of Open CV is written in C++. There are now full interfaces in Python, Java and MATLAB. Wrappers in other languages like C#, Perl and Ruby are developed. A CUDA-based GPU interface has been ongoing since 2010. Open CV has in received support from Intel and recently it's received support from willow garage, a privately funded new robotic research institute. Opens can run on different platforms like windows, android.

The rapid detection is finished by employing a cascade classifier (fig. 2). The image is segmented in sub-windows and only some features are checked at the start.

B. Face Recognition

It is another phase, where the detected face image is compared with images within the database of faces. The opens framework contains the inbuilt face detector that may work 90-95% on the clear images. It’s slightly difficult to detect a face if an individual wearing glasses or a picture is blurring.

In fig. 3 an example image is given in a very. This image is already screened in sections. Our application for benchmarking will display a colored image sequence but analyze only the desiderated grayscale image.

Fig. 1. Haar-like features

Fig. 2. Detection cascade T=true; F=false

Fig. 3. LBPH analysis
II. ALGORITHM IMPLEMENTATION

Before a benchmarking of the Open CV algorithms can be performed, software had to be developed which implements the face detection and recognition algorithm. Implementing the popularity algorithm requires to also implement Open CV’s face trainer. This led to whole software which had to be developed with the subsequent features:

1. Read within the video stream from a camera which is connected to the pc.
2. Detect a face within the current frame of the image sequence

A graphical representation of the implemented features and software architecture are often seen in fig. 4.

Fig. 4. Software Architecture

III. IMPLEMENTATION DETAILS REGARDING SPEED

The complete software was written in Python and executed in a very Python V.2.7.11 environment. The Open CV algorithms are written in C++ and were called using the Python application. The timing overhead which occurs when calling the libraries from within the Python application are often neglect, additionally all the algorithms which are benchmarked are written in C++ and were compiled to binaries for the particular system.

IV. FACE DETECTION

Generally developing applications which detect a face is incredibly easy when using Open CV. The detection itself is simply one line of code; the” configuration” of the face detector is additionally quite lightweight. In line two the file with the haar-cascades is loaded into the RAM disk. This can be an XML based file which holds pre-trained information necessary to detect an individual's face. Loading another cascade classifier ends up in detecting other objects, as an example a dog’s face.

V. BENCHMARKING

As already described in section II-D the face detection and also the face recognition is essentially performed with only 1 command. Therefore, the benchmarking consists of measuring the execution time of the particular command which starts the detection and recognition. Measuring the execution time can programmatically be performed in simple steps:

1) Get current time (by asking the OS) and store it as start time
2) Run the function or method whose execution time should be determined.

VI. PERFORMING THE BENCHMARK

Preconditions

When performing the benchmark several important preconditions had to be ensured:

- Illumination conditions need to be the identical for training the recognizer and for running the recognizer
- Training the recognizer was done from live camera stream

Benchmarking on devices

To get a control about the execution time of the face detector and recognizer the concept was to run the identical tests on several devices. This shows...
whether the algorithms support multi core systems (“increasing number of cores, the faster the execution becomes.”

VII. RESULTS

The benchmarking of the applying was done separately for the detection similarly as for the popularity phase. For all recognizer test cases in table II the camera resolution was specified to 800*600 pixels and also the samples of two persons were used (fig. 7). The identical camera resolution was used for the face detector.

VIII. CONCLUSION

The computational models, which were implemented in this project were chosen after extensive research and the successful testing results confirm that the choices made by the researcher were reliable. The system with manual face detection and automatic face recognition did not have a recognition accuracy over 70% due to the limited number of eigen faces that were used for the PCA transform. This system was tested under very robust conditions in this experimental study and it is envisaged that real-world performance will be far more accurate.

In this we present a collection of Python biometric performance benchmark algorithms called OpenCv. The algorithms are fast enough to be useful in realtime systems however, improving performance would allow the algorithms to process more images.

The fully automated frontal view face detection system displayed virtually perfect accuracy and in the researcher’s opinion further work can be conducted in this area for increased performance and accuracy.

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