



Design and implementation of Augmented Reality learning system using contour analysis

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Abstract— Augmented reality (AR) has been increasingly applied in various fields. It is a new technique of the computer vision application used to facilitate interaction in the digital arts. Augmented Reality (AR) employs computer vision, image processing and computer graphics techniques to merge digital content into the real world. It enables real time interaction between the user, real objects and virtual objects. This paper propose different analysis methods for processing an image with more emphasis given on contour analysis. As contour contains the necessary information on the object shape, interior points of object are not accepted to attention. That lowers the computing and algorithmic complexity. We learn the different patterns to the system, and then augmenting the assigned virtual object to it. It is an advance method and could be a promising technology for motivating users to engage in learning systems.

Index Terms— Augmented reality, Pattern recognition, Image processing, Contour analysis.

I. INTRODUCTION

Augmented Reality aims at simplifying the user's life by bringing virtual information generated by computer not only to his immediate surroundings, but also to any indirect view of the real-world environment, such as live video stream. The objective of this research is to provide tool to learn different patterns that will be inputted into the system. And augment the assigned object to these patterns. By doing this we will be able to enhance the user perception for the respective pattern. The first part of the article contains reviews of definitions and algorithms in augmented reality, image enhancement, edge detection and image segmentation as in [1]. These algorithms are applied to use on our problem domain where we try to augment the learning image. The second part is the proposed model that designed to extract the image by using inbuilt library and close contour of small elements of the characters. Third part includes augmentation.

II. AUGMENTED REALITY

AR is a type of virtuality which aims to duplicate the world's environment in a computer. AR generates a composite view for user which combines real scene viewed by user and the virtual scene generated by computer that augment the scene with additional information. Encyclopaedia Britannica gives the following definition for AR: "Augmented reality, in computer programming, a process of combining or 'augmenting' video or photographic displays by overlaying the images with useful computer-generated data."

Simple augmented reality system as shown in figure 1 consists of a camera, a computational unit and a display. The camera captures an image, and then the system augments virtual objects on top of the image and displays the result as in [4]. The capturing module captures the image from the camera. The tracking module calculates the correct location and orientation for virtual overlay. The rendering module combines the original image and the virtual components using the calculated pose and then renders the augmented image on the display.



Figure 1. Example of a simple augmented reality system setup.

The tracking module is the main part of the augmented reality system, it calculates the relative pose of the camera in real time.

III. DIFFERENT TRACKING METHODS

Marker-based tracking

As in most augmented reality setups the camera is already part of the system, visual tracking methods are used in AR. In visual tracking, easily detectable predefined sign is added in the environment and computer vision techniques are used to detect it. A marker is such a sign or image that a computer system can detect from a video image using image processing, pattern recognition as in [8] and computer vision techniques. Once detected, it then defines both the correct scale and pose of the camera. This approach is called marker-based tracking as in [16], and it is widely used in AR as in [5]. Feature-based tracking

Feature detection and tracking algorithms are widely used for motion detection, image matching. We can divide localised features into three categories: feature points (e.g. corners), feature descriptors (e.g. SIFT) and edges. A feature point is a small area in an image, which has a clear definition and a well-defined position.

Visual tracking method

A visual tracking method as in [9] deduces the camera's pose from what it sees; therefore, visual tracking is often called camera(-based) tracking or optical tracking. Visual tracking can be based on detecting salient features in the images, this approach is called feature-based tracking as in [3] and [7]. The system may also have a model of the scene or part of the scene and then tries to detect this model from the image and thus deduce the pose of the camera; this approach is model-based tracking.

Feature detection and tracking algorithms are classified based on what kind of features they detect: edge detectors (e.g. Canny), corner detectors (e.g. Shi&Tomasi), blob detectors (e.g. MSER) and patch detectors ..

IV. FEATURE MATCHING

After detecting the features, the system needs to match them, i.e. it needs to find corresponding features in different images. For feature matching, tracking systems use commonly two different approaches: they compare the small image areas around the features and find similar areas (template matching) as in [2], or they calculate image characteristics around the features and compare them (descriptor matching).

A) Scale-invariant feature transform (SIFT)

It is a widely used feature detection and tracking algorithm. SIFT is based on feature descriptors. The SIFT algorithm computes a histogram of local oriented gradients around the interest point and stores the bins in a 128-dimensional vector (eight orientation bins for each of the 4×4 location bins).

B) PCA-SIFT

It is a variation of a SIFT algorithm which is also based on the salient aspects of the image gradient in the feature point's neighbourhood. PCASIFT applies principal components analysis (PCA) as in [13] to the normalised gradient patch image instead of using SIFT's smoothed weighted histogram. PCA-SIFT yields a 36-dimensional descriptor which is faster for matching, but has proved to be less distinctive than SIFT in the performance evaluation test .

C) Gradient Location and Orientation Histogram (GLOH)

It is also a SIFT-like descriptor that considers more spatial regions for the histograms. GLOH uses principal components analysis like PCA-SIFT, but yields to a 64-dimensional descriptor .

D) Speeded Up Robust Features (SURF)

It is a scale and rotation-invariant feature point detector and descriptor for image matching and object recognition [2]. SURF as in [14] is based on sums of 2D Haar wavelet responses and makes efficient use of integral images. As basic image features, it uses a Haar wavelet approximation of the determinant of Hessian blob detector. The standard version of SURF is faster than SIFT and more robust against different image transformations than SIFT [3].

E) Local Energy-based Shape Histogram (LESH)

It is a robust front-end pose classification and estimation procedure originally developed for face recognition. It is a scale-invariant image descriptor, which can be used to get a description of the underlying shape. LESH features suit a variety of applications such as shape-based image retrieval, object detection, pose estimation, etc. LESH is based on a local energy model of feature perception. LESH accumulates the local energy of the underlying signal along several filter orientations, and several local histograms from different parts of the image patch are generated and concatenated together into a 128-dimensional compact spatial histogram.

F) Contour analysis (CA)

If the necessary information is in the object shape, we can go for CA. In a CA the contour is encoded by the sequence consisting of complex numbers. It allows to effectively solve the main problems of a pattern recognition - transposition, turn and a rescaling of the image of object. CA methods are invariant to these transformations.

V. CONTOUR ANALYSIS

The contour is the boundary of object. In a CA the contour is encoded by the sequence consisting of complex numbers. On a contour, the starting point is fixed. Then,

the contour is scanned and each vector of offset is noted by a complex number $a+ib$. Where a - point offset on x axis, and b - offset on y axis. Offset is noted concerning the previous point. Owing to the physical nature of three-dimensional objects, their contours are always closed and cannot have self-intersection. Hence we can define unambiguously a way of bypass of a contour. The last vector of a contour always leads to the starting point. Each vector of a contour name(EV).

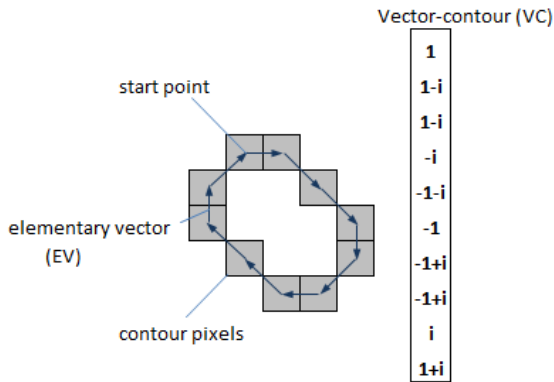


Figure2. CA analysis

And sequence of complex-valued numbers - vector-contour (VC). Normalised scalar product (NSP) of a contour can be defined as

$$\eta = \frac{(\Gamma, N)}{|\Gamma||N|}$$

The norm of the normalized scalar product of contours gives unity only in the event that these two contours are equal to within turn and a scale. Otherwise, the norm of NSP it will be less unity. Actually, the norm a NSP is an invariant on transposition, rotation and scaling of contours. If there are two identical contours their NSP always gives a unity, is not dependent on where contours are, what their angle of rotation and a scale. Similarly, if contours are various, their NSP will be strict less 1, and also independent of a place, rotation and a scale.

If contours are identical, but the EV reference begins with other starting point the norm the NSP of such contours will not be equal to a unity.

VI. PRACTICAL APPLICATION OF CA

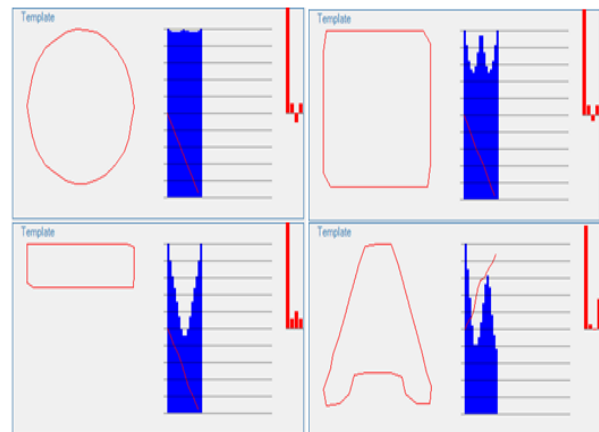
Contour analysis is used for pattern recognition task as in [17] on the image. Let us take the image a size $n*n$ pixels. Then breed its uniform grid with a step s . The total length of all grid lines is:

$$L = \frac{2n^2}{s}$$

The CA allows to process the image in a progressive mode. It means that we can sort contours on any to an indication (for example, by square or on a gradient of boundaries, or on brightness, etc.). And then to treat the first contour, and to produce outcome. Remaining contours to process in a background mode. As contours are independent from each other algorithms of a recognition it is easy to parallelize. Besides, algorithms are very simple and can be executed on graphic processors.

For fast searching of templates, it is necessary to introduce the certain descriptor characterizing the shape of a contour. An ACF invariantly to transposition, rotation, scaling and a starting point choice. And besides, the ACF is a function of one contour, instead of two, as an ICF. Hence the ACF can be selected as the descriptor of shape of a contour. The close contours will always have the close values an ACF.

In pictures, the norm the ACF is represented by dark blue color (an ACF it is represented only for an interval from 0 to $k/2$).



Figur3.ACF for different shapes

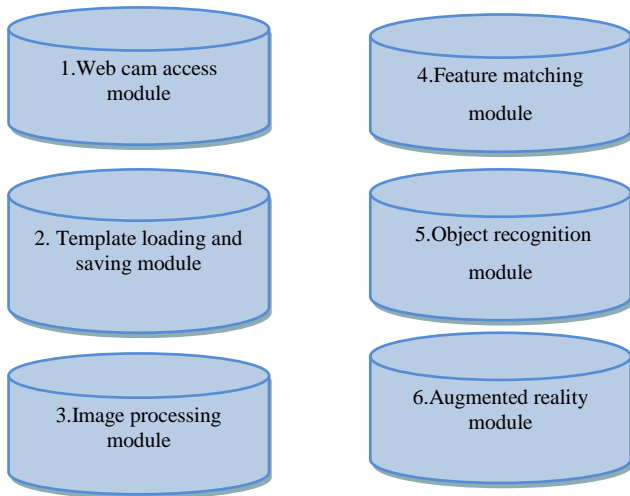
In the real image contours have arbitrary length. Therefore, for searching and comparing of contours, all of them should be led to uniform length. This process is called equalization.

VII. LEARNING SYSTEM

1. Web cam access module

It consists of program to access web camera in order to take input from the web cam. Here C#.net and openCV software is used to access camera. As soon as the software starts, this module of the software starts executing. By clicking on start button in GUI learner can allow the web cam to take input. Output of this shows "camera started". By clicking on stop button in GUI learner can allow the

web cam to stop processing. Output of this shows “camera stopped”.



Figur 4. Different modules

2. Template loading and saving module

This module fetches previously stored or learned templates from the database. Database consists of table in which augmented picture for particular template is there. Contour templates with assigned pictures are stored in table form. When a given pattern is matched with the contour template, assigned picture is display for that contour. User can store new template by clicking on learn template button in GUI.

In learn template module, we can learn new templates to the system. The contours for given real time images are detected first, and particular pattern as a template is selected. We can assign new jpg or png picture to that particular pattern, by clicking on add selected contour as template button.

3. Image processing module

It takes the image from the camera module and processes it. It involves smoothing, dialation, erosion and thresholding etc.

4. Feature matching module

After detecting the features, the system needs to match them, i.e. it needs to find corresponding features in different images. For feature matching, tracking systems use commonly two different approaches: they compare the small image areas around the features and find similar areas (template matching), or they calculate image characteristics around the features and compare them (descriptor matching).

In the template matching approach, a feature detector matches image patches against regions in the image. This

is also called the patch matching approach. Template markers and image markers can be considered image patches, with the distinction that marker detection is a separate process and matching is used only for identification of the marker.

The second matching approach where the system calculates image characteristics around the feature that describe it appearance distinctively. These calculated characteristics are called feature descriptors.

A good feature descriptor is invariant under image deformations, it is scale and illumination invariant and is capable of distinguishing between different features.

An ACF invariant to transposition, rotation, scaling and a starting point choice. And besides, the ACF is a function of one contour, instead of two, as an ICF. Hence the ACF can be selected as the descriptor of shape of a contour. The close contours will always have the close values an ACF.

This module contains set of codes matching features of the recognized object/image which is pointed out by the learner and image in the database.

5. Image/Object recognition module

It contains code for Image/Object recognition. Image/Object recognition is done by the following procedure. When we have all pairs of matched features, we can recognize the images or objects. The simplest method is to count the number of matched features. Without loss of generality, the image pairs that have the largest number of matched features are the same.

6. Augmented reality module

In this database all augmented reality shapes which has to be augmented out when certain condition is matched ,for example when images are matched is stored. This is also updated whenever a new pattern is encounter. These augmented reality patterns are changes from Images to images / Objects to Objects matches. When the camera is focused on the particular character on paper the student can study that by watching augmented reality pattern.

VIII. CA LIMITATION

CA makes sense, only in that case when the object contour is defined unambiguously correctly in all points.CA methods assume that the contour describes all object bodily, and does not suppose any intersections with other objects or incomplete visibility of object.

IX. SOFTWARE

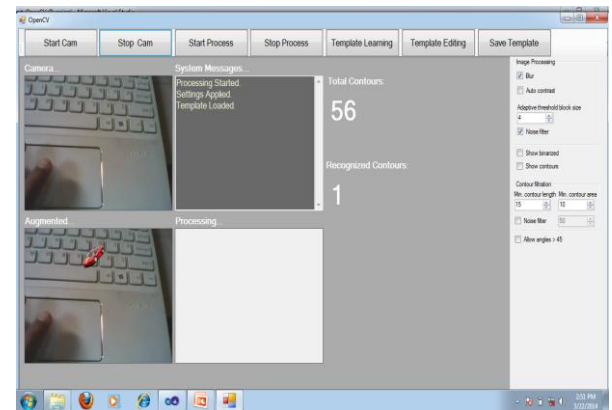
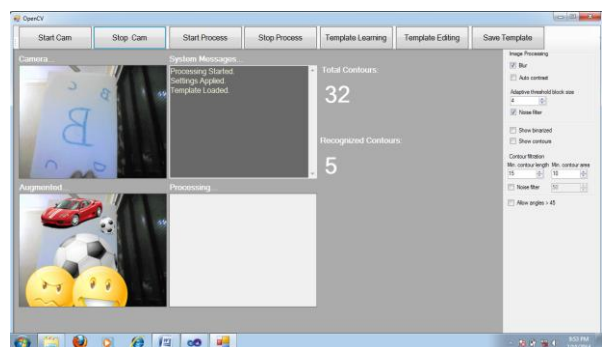
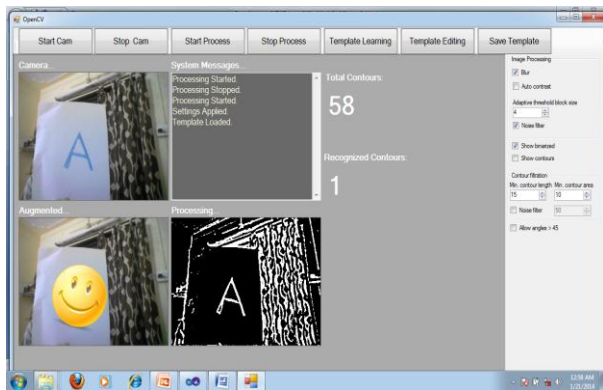
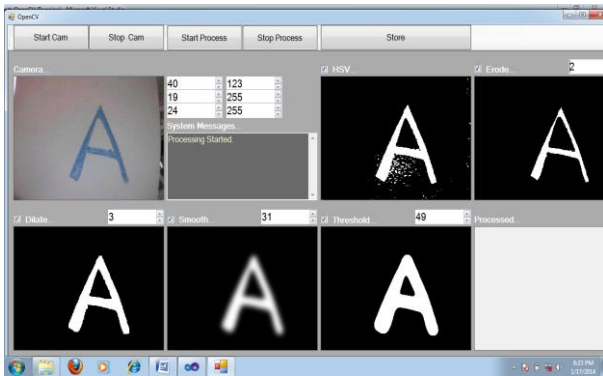
Some software's for image processing are – matlab which is very slow in response, AForge.net which don't have more libraries for human computer interaction (HCI) and OpenCV. OpenCV as in [18] is an open source computer vision library written in C and C++ and runs under Linux, Windows and Mac OS X. Image going to be processed is capture through web camera using open CV software.

As C#.net is based on C++ and is compatible with OpenCV library. For implementing base function of CA we process C#.

Contour Analysis Processing uses library OpenCV (EmguCV .NET wrapper) for operation with the image. C# is easy to use and is graphical user interface (G.U.I.) based.

Emgu.CV is used as a wrapper for the C#.net and OpenCV environment.

X. RESULTS



XI. ADVANTAGES

- It helps students to understand and the concepts clearly.
- The chance of false imagination while learning is reduced.
- Less time to learn.
- Recall of concepts is enhanced.
- Self learning is possible or it is easy for lecturers to teach.
- It can be used in understanding geometrical puzzles, Structure of electron, atom...etc, g)
- It leads to innovative thinking
- It improves the standard of education.

XII. CONCLUSION

CA methods are attractive for the simplicity and high-speed performance. high-speed performance of a CA allows to process video.

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