

Image Extraction By Human Motion Detection Using 3D Computer Stereo Vision

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Abstract

Computer stereo vision is the extraction of 3D information from digital images, such as obtained by a CCD camera. By comparing information about a scene from two vintage points, 3D information can be extracted by examination of the relative positions of objects in the two panels. This is similar to the biological process Stereopsis. The system captures monocular video of a scene and identifies those moving objects which are characteristically human. This serves as both a proof-of-concept and a verification of other existing algorithms for human motion detection.

Keywords—CCD camera; Stereopsis; monocular video; human motion detection.

1. Introduction

The initiated motion of human motion detection is being done on the basis of tracing a mapped view of whole human body based motions [1]. This is governed by the SVS decomposition of point-to-point human motion detection in terms of varied motion and posture of human body [2]. Therefore, it provides an in depth research to obtain the human motion and its preferred estimation as to in calculate the exact visual platform [3].

However, this paper provides the use of CCD camera which compares the image from two different arbitrary points developed with SVS in reference to principal of imitation. As considered for the first perception the background is detected with level of its image slicing required for providing an image based

threshold value [4]. Concern to this is the Image to Noise ratio in the binary image which is enhanced in contrast filtration by morphological filtering. Image to Noise ratio is maximized with reduction of noise and a prefaced blob of target is been considered. Human motion detection has five features of postures in point reference (i.e., head, four tips of two arms and legs) are evaluated in estimation to central segmentation of moment and characteristic processing of these five features provides this paper with an specified proof of understanding with evaluation of human based SVS motion in terms of CCD camera with suitable comparing between the image points in reference to its panel segmentations[5]. It leads to extraction of 3D Stereo Vision of human motion in respective frames in terms of its background registered for motion detection.

2. Setup and Task Description

A. Setup Considered

First, considering the image imitator in terms of CCD camera unit is proposed in terms of stereo vision system (SVS) for a controlled pulse-width signal modulation (PWM).confirm that you have the correct template for your paper size. This IS considered for X-Y plane corresponding field of view (FOV) for respective planes confined to horizontal and vertical planes. A window for frames representing (X_1, Y_1) , (X_2, Y_2) and lastly (X_3, Y_3) for its specified distances respectively is confined to frames of larger FOV [6].

B. Task Description

The first task is to define in terms of background of sequence of 3-D motions detected by specified image ratio with background registration [7]. This is considered only after the image based threshold and morphological filtering. This leads to the estimation of five feature based points of human body needed for video extraction by 3-D motion [8].

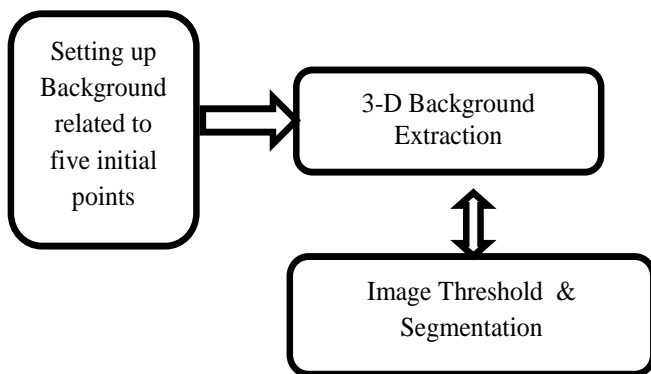


Fig. 1. Block diagram of Image Extraction in terms of Human detection

Considering, as $L_n(i)$ is the line linking from $a_n(i_{min})$ to $a_n(i_{max})$, where $i_{min} = \min\{r_n(x,y)\}$ and $i_{max} = \max\{r_n(x,y)\}$. Provided that perpendicular distance from $a_n(i)$ to $L_n(i)$ having $D_n(i)$ in terms of binary image obtained as follows:

$$b_n(x,y) = \begin{cases} 1, & \text{if } r_n(x,y) > i_T \\ 0, & \text{if } r_n(x,y) \leq i_T \end{cases}$$

Where $i_T = \arg \max \{ D_n(i) \}$

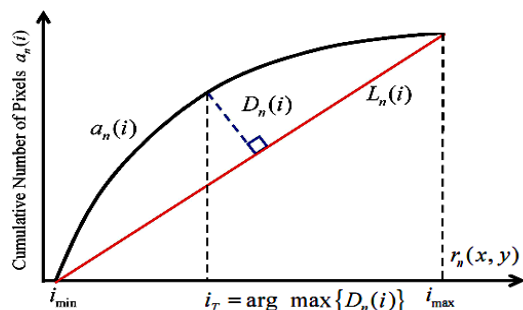


Fig 2. Threshold point of histogram in terms of change of point [1]

This provides the background in terms of its coordinate values of (x,y) which are thus

represented as for $b_n(x,y) = 0$ are accumulated for $s_n(x,y)$ zero initial value:

$$s_m(x,y) = \begin{cases} s_m(x,y) + 1, & b_n(x,y) = 0 \\ 0, & \text{otherwise for } 1 \end{cases}$$

Where we have $n - n_f \leq m \leq n$. A pixel is thus masked in terms of its stationary values of n_f for all successive frames. Perspective threshold value of the cumulated histogram curve provided in Fig 2. is obtained in a manner for obtaining the dependent nature of its frames [9] [10].

3. Posture Estimation

The central moment can be represented with rotation axis having minimum inertia of object. Hence, it is applied to calculate the direction of silhouette of human. First, the center of gravity (COG) of silhouette for human is denoted as (x_c, y_c) given by

$$x_c = \sum x / N, y_c = \sum y / N$$

Where x and y are the pixels in respect to the various frames to be obtained in reference to its background. The nearest point to its detection is based on the phase shift of ϕ_c is set from an range of minimum valued 30^0 to maximized shift of 150^0 as after that there is transition in the image contrast and its threshold is lost [11].

There will be an difference in the frames of detection such as to consider all the frames as of the form D_1 to D_{K-1} such that above frame pattern as shown in Fig 3. will be detected [1] [9].

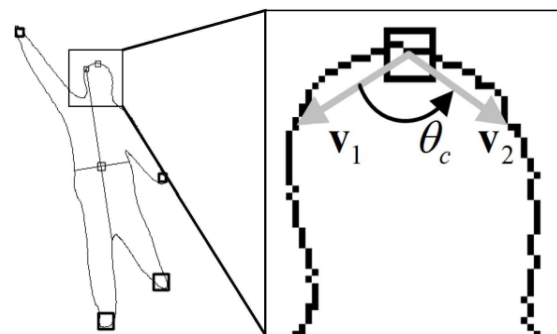


Fig 3. Illustration of ϕ_c of head for human motion

Various frames changed with $Smg(k)$ denotes the state of moving index at the k th feature point(s) are the candidates of key-posture frames. On the other hand, the state $Smg(k) = 0$ indicates that all the five features of human body motion remains unchanged. Meanwhile, considering analysis, of rising signal of $Smg(k - 1)$, i.e., $Smg(k - 1) = 0$ and $Smg(k - 1) = 1$, is satisfied, then $Smg(k - 1)$ is assigned as key-posture frame, i.e., $Sk(k - 1) = 1$, indicating one key-posture frame is found at the $(k-1)$ th frame.

4. Expected Result

Previous, motions considered with respect to present extraction at its frontend side with 3-D motions, with coordinates of these five features are obtained by the suggested SVS [2] [5]. They are then fed to the experimental setup of CCD camera to thus, proceed for image extraction with respect to an matrix based segmentation.

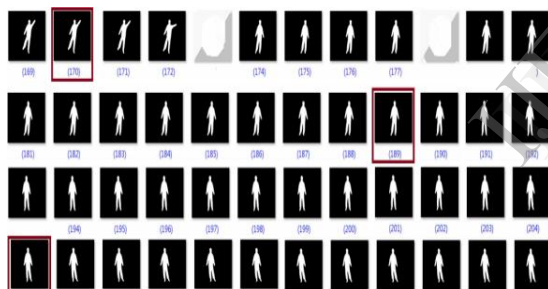


Fig 4. Matrix enabled background based Segmentation and Threshold [1].

5. Expected Conclusion

As per the proposed human motion with a CCD camera enabled stereo vision system (SVS) captures a sequence of 3-D motion images of a human, which is faced to the proposed HR. After the inquiry of enough motion sequences, the motion detection via image ratio and background registration, the estimation of five feature points (i.e., head, four tips of two arms and legs) of a human, are obtained. Based on this human body with five

estimated feature points, the key-posture frames from the video of 3-D motion of a human are extracted. The corresponding experimental results confirm the effectiveness of the proposed novel method. Finally, they are applied to imitate the 3-D motion of a human body based detection in terms of real time systems.

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