The Study of Monocular-Vision-Based on Moving Object

Xi Lin¹, Liu Dong¹ *

¹ Beijing Union University, Beijing 100101, China

Abstract. In order to satisfy the moving object diction and tracking’s real-time and self-learning performance, the method which combines background substraction and threshold image segmentation on HSV space is presented in this paper. Detecting the real-time image and getting the threshold of HS space. Next, the moving object’s area is got by the thresholds segment. Third, getting the center and realize the vision-tracking of moving object by the robot servo-experiment.

Keywords: Vision-tracking, Background substraction, HSV.

1. Introduction

This paper deals with the problem of detection and tracking a moving object via a monocular-vision mobile robot through visual feedback. Detection (recognizing a moving object which is breaking into the field of robot-vision) and tracking (approaching a moving object until a stated distance) are important tasks in a number of applications, ranging from robotic games to automated surveillance. Moreover, development of effective methods for performing these tasks represents a challenging tested for the integration of various techniques involving image processing, filtering, control theory and artificial intelligence (AI) strategies. This thesis focuses on how to get the moving object character efficiently and ensure the real-time tracking is very important[1].

In this literature, object tracking in an image sequence using an image segmentation that is based on getting the basic image features of the moving object, which break into an static background in this paper. The mobile robot uses Monocular-Vision system which is structured by vehicle-holding CCD vidicon. This structure makes the vidicon can move with the mobile robot, just like a human being decides his action by what he saw in eyes. And it not only has better adaptability but also makes coordinates can much more easily be conversed. The vidicon fixed it’s focus for getting clear image. Because the stability of vehicle-holding CCD vidicon is not very good, in the experiment mobile robot is moving on smooth ground. For the demands of satisfy real-time and self-learning performance in the study on tracking moving object, this literature combines background substraction and edge detecting to getting the HSV thresholds of the moving object’ color feature, and then using HSV space based threshold segmentation distinguish the moving object from the real-time image sequence to achieve moving object tracking.

2. Getting moving object feature

2.1. Moving object detection

Background substraction is a usual method to detect moving object in image processing. It uses the differences between the current image and the background image to detect the moving object’s area in current image. Because it studies detecting a color moving object break into an static background in this paper, getting a steady-going background image is realizable. So using background substraction to detect moving object can get an entire object’s image as follows:

* Corresponding author. Tel.: +86-01064900225; fax: +86-01064900850.
E-mail address: zdhtliudong@buu.edu.cn.
Firstly, a background image should be stored which captured by vidicon. And a background image statistical model \( B(x,y) \) will be founded by every pixel in the image. Secondly, the current k-th frame image \( f_k(x,y) \) subtracts the background image \( B(x,y) \), and computing the pixel of \( f_k(x,y) \) which departure the value of \( B(x,y) \) more than the threshold \( T \). And these pixels are judged in the moving object, as follow formula:

\[
D_k(x,y) = \begin{cases} 1, & |f_k(x,y) - B_{k-1}(x,y)| > T \\ 0, & \text{others} \end{cases}
\]

\( T \) is grey threshold value. The detected area of object are composed by the pixels of \( f_k(x,y) \) which computing values of \( D \) equal 1.

Background substraction can efficiently detect if there is a moving object in current image frame, but can not identify if the moving object is enter the image entirely. So this arithmetic is improved by combining with an object location identify method in this paper. On the assumption, current image frame \( f_k(x,y) \) object includes \( N \) points which recorded as \( P_i = \{ (x_i,y_i) | i = 0, 1, 2, 3, \ldots, N-1 \} \).

If it doesn’t exist a point \( P_i = \{ (x_i,y_i) | (x_i = 320 \cup y_i = 240) \} \), \( i = 0, 1, 2, 3, \ldots, N-1 \), it judges that the object was enter the image entirely in the k-th image frame, and the object area we got in this image were integrated.

The experiment proved that the background substraction combines with object location identify method can avoid losing information, for not gathering moving object as information of moving object while the object enter the image entirely.

### 2.2. Getting moving object’s HSV threshold values

When the object area is certain, this area which is RGB image should be converted to HSV form and get the HSV threshold values. This is because the robot uses RGB-color-based vision system in this study. But some factors, which brought by vidicon like high-lightness, shadow and so on, will affect image processing crucially. The RGB color space is not efficient in describing objects’ color and dealing with calculation. So this literature uses another color space – HSV to process the color image. HSV model is much more like what human apperceived of color. Its hue attribute can reflect the color category exactly and be less affected by the environment’s illumination and has stable and narrow values range, so can make it to be one of the primary parameters. Compared with RGB model HSV model can be the basis of detection better[3-4]. This paper uses HS model, which predigests from HSV system, to get the object color feature threshold values. Because it can certain the object’s color attribute and independent from lightness information. Comparing with HSV model, it is not only reduce the calculation but also assure the veracity, advancing the capability of real-time image processing. The formula that convert RGB model to HS model is as follows:

\[
\begin{align*}
H_0 &= 0 \\
G_0 &= G - B \\
R_0 &= R - B \\
H &= H_0 + 120(G_0 + R_0) / (max + min) \\
S &= (max + min) / min
\end{align*}
\]

The max and min are the max and min of original RGB image’s three components’ \( R;G;B \), then obtain the histogram of the moving object and certain the HSV image’s thresholds \( H_{min};H_{max};S_{min} \).

### 3. HSV-threshed-based segment of moving object tracking

Manuscript is accepted for review with the understanding that no substantial portion of the paper has been published or is under consideration for publication elsewhere and that its submission for publication has been approved by all of the authors and by the institution where the work was carried out. It is further understood that any person cited as a source of personal communications has approved such citation. Articles and any other material published in the proceeding represent the opinions of the authors and should not be construed to reflect the opinions of the Editor(s) or the Publisher.
Firstly, the RGB images real-time gathered by image grabbing card should be converted to HSV images. Then the image compare with HSV model’s threshold values. The pixels are moving object’s, when the comparisons satisfy follow conditions:

\[ H_{min} < H < H_{max} \]
\[ S_{min} < S < 1 \]

Then the background can be segmented from the image, and getting the moving object.

After HSV threshold segment, the color image is converted into two valued image. Then the object’s contour is got by borderline tracking arithmetic. On the assumption, the borderline includes M points in the image, recorded as \( S_i = \{ (x_i, y_i) | i = 0, 1, 2, 3, ..., M-1 \} \). The \( (x_i, y_i) \) is the i-th point’s coordinates. And the centroid of the object’s area is defined as \( P_c = (X_c, Y_c) \), as follows are the formulas to get centroid:

\[
P_c(X_c, Y_c) = \left\{ \begin{array}{ll}
X_c = \frac{1}{N} \sum_{i=0}^{N-1} x_i \\
Y_c = \frac{1}{N} \sum_{i=0}^{N-1} y_i
\end{array} \right.
\]

The experiment proved that the HSV-threshold-based image segment can get robust segment results during different illumination conditions when light intensity is not varies greatly, And the result is good, even the object is colorized, nonsolid color or parts of the object is kept out. So this method, using HSV-threshold-based image segment to get the object’s borderline, can get good result in this paper.

4. Experiment and simulation

Before the tracking begins, a digital RGB image should be received firstly as the background image from image grabbing card. When moving object moves into the CCD vidicon’s visual field, an area of the object will be parted from the real-time image with background subtraction. If the object area pass the object area detection, this area will be converted into HSV form, and a HS histogram will be built. The \( H \) and \( S \)’s thresholds of the moving object can obtain from the histogram. This processing corresponds to a self-learning course. During the following tracking course, the thresholds’ standard will not change until a next new tracking causing a new learning. Then begin track. Comparing the real-time image, which is captured by image grabbing card, with obtained \( H \) and \( S \)’s threshold values, a centroid of the moving object that is parted from background will obtain (Fig. 1). Figure 1(a) is background image. Figure 1(b) is real-time image. Figure 1(c) is the result getting through background substraction, and it is judged that the object does not enter the visual field entirely by object area detection. So next image frame (Fig. 1 (d)) is took. Figure 1(d) pass the detection, then obtain it’s HS threshold values. Figure 1(e) is a two valued image converted from Figure 1(d). Figure 1(f) is the centroid.

The size of tracking interface is 320×240 pixels. In this experiment, the relation between the moving object’s centroid (the moving object’s ubiety relate to the robot) and robot’s rate and angle’s should be obtained in advanced by existing knowledge and experiences. Figure 2 is the object tracking image. The dots of the picture are the object’s centroids moving track. It was based on the detected moving object’s centroids in every image frame of the image sequence. It can be seen that most of the dots distributed in this area of tracking interface: X-coordinate (80~240) and Y-coordinate (60~180), excepting the rate of moving object change intensely.

This experiment proved that the proposed method in this paper can efficiently track the solid color or nonsolid color object’s motion. It can be seen that this method can make robot taking a quite stable tracking result in short time, for most of the moving object’s centroids are in the middle part of the visual image.
Conclusions:

In this literature, the method combines background subtraction and moving object detection are to tracking the moving object which breaks into the robot’s visual field, and self-learning the color feature of object. The experiment not only shows that the method can obtain the feature information of object completely, but also embodies the Self-adaptive and self-learning properties of mobile robot during the object motion tracking. And using the HSV-space-based HS threshold values as the object feature can reduce the affection from some factors like: light, shadow, nonsolid color and so on. This method efficiently realizes moving color-object tracking.

5. References


