Night Time Vehicle Detection Using Tail Lights: A Survey

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Abstract- The main challenge during night-time driving is the visibility of the road ahead. The major challenge rests in controlling accidents during night as the night conditions differ from the day in environment conditions, vehicle lighting etc. The reduced cost of cameras and optical devices helped in installing front-mounted intelligent systems for identifying forward collision avoidance and mitigation. During night, vehicles in front are generally seen by their tail lights. The turn signals are mostly important because they signal lane change and potential collision. This survey presents a car recognition system by identifying and segmenting tail lights in the night-time road environment.

Keywords

Segmentation, Edge detection, Taillight, Blob detection, Connected component analysis, Symmetry score, Nakagami-m distribution .

INTRODUCTION

This paper gives a literature review on various techniques used in identifying vehicles during night. The usage of Computer Vision ideas has helped lot with road safety and security. The appearance of vehicles during night time is different when compared to its daylight conditions such as environment lighting, reflection of light on the body of vehicles, color of vehicles etc. Thus a different image processing approach is essential for night time road environment. This method seeks on segmenting the tail lights of vehicles and classifying them. When vehicles are viewed from behind at night, they are primarily visible by the red color of tail lights. All car models have their own peculiar physical and structural features which make them distinctive from each other and the appearance of brake lights is one of those features. For the technique to be most efficient, ideally a tweaked camera is used which filters only the red lights, thereby eliminating image bleeding.



Fig 1 Tail light of vehicle



Brake light of Vehicle

TECHNIQUES

Segmentation

Segmentation is the process by which an image is fenced off into smaller parts so that processing of an entire image can be more meaningful and easier i.e. process of partitioning a digital image into set of pixels. The process of segmentation makes an image meaningful and easy for analysis and it assigns label for every pixel so that same characteristics are shared by pixels with same label.

BhavinkumarM.Rohit et.al. [1] used the segmentation approach and introduced low light video frames which have low exposure value images. The idea behind using the low exposure image frames is that factors such as street lights, unwanted reflections from vehicles

body, sign boards reflections can be removed and only the bright red color and head lights of oncoming vehicles are visible in the image frame. Segmenting the red color from other noises is the major issue in identifying the tail lights.

Binamrata Baral et.al. [2] explains Special Theory Based Segmentation techniques such as Genetic Algorithm, Neural Network, Clustering and Wavelets based segmeentation techniques. Genetic algorithm depends upon the process of natural selection and uses techniques such as selection, cross over and mutation. In neural based, the principle of ANN is used. For clustering Fuzzy C-means and K-means clustering is implemented and in wavelet based wavelet transforms of image is used for feature extraction.

JunbinGuoet.al. [3] explained OTSU algorithm for tail light segmentation for identification of vehicles at night. In this after analyzing the histogram of image the lower boundary for thresholding is determined by counting the average number of higher gray scales. Next, between the lower boundary and the highest gray scale of the image, the optimal threshold is calculated by Otsu method and the image is divided into pixels based on this threshold. The resultant object I_{result} can be computed by,

$$I_{result}(x,y) = \begin{cases} 1 & \varphi_{\emptyset RC}(x,y) > T \\ 0 & \varphi_{\emptyset RC}(x,y) < T \end{cases}$$
(1)

where $\varphi_{\phi RC}$ is the image and T the threshold value.



Fig 2 Histogram showing apparent classes using thresholds T_1 and T_2

Edge detection

The edges of an image indicates higher frequency region within that image.Edge detection refers to the process of identifying and locating abrupt changes in pixel intensity which characterize boundaries of an image.Detection of edges is applicable in image segmentation, data compression image reconstruction and so on. There are two types of edge detection; first order and second order edge detection. The techniques of edge detection are Sobel, Prewitt and Canny edge detection.

P.Srinivaset.al.[4] implemented Canny edge detection in order to detect the vehicle tail light. The algorithm smoothens the image and finds the gradient to highlight regions with high spatial derivatives. The non-maxima regions are then suppressed and two threshold values are set. The value is set to zeroif the magnitude is below the first threshold, it is considered as edge if the magnitude is above second thresholdand a path is obtained if magnitude lies between the two.

M.Kalpanaet.al.[5]explained edge detection operators such as differential operator like Robert, Prewitt and Sobel operators, Log operator and Canny operator. Among these operators the most commonly used is the Canny operator. Even after edge detection various noises exists in the detected image and then wavelet transformation has been used to remove that.

Manoj K Vairalkaret.al.[6] proposed Sobel edge detection for image detection. The Sobel detector is selected after verifying the edge orientation, noise environment and edge structure. Depending upon the output of these variables the techniques implemented for edge detection are gradient and laplacian detection. The Sobel operator is associated with a pair of 3x3 convolution kernel, one obtained by rotating the other one at 90° . The gradient magnitude G is given by,

$$G| = \sqrt{\left(G_x^2 + G_y^2\right)} \tag{2}$$

where G_x and G_y are magnitude in x and y direction respectively.

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Filtering and Enhancement

Filtering is the process of modifying or enhancing an image by removing the noises from it and is a neighborhood operation in which the output image pixel value is obtained by applying some algorithms to the neighborhood pixel values of the input pixel. Image enhancement is the process by which the specific features of an image is brought out by histogram equalization, median filtering *etc*. Enhancement makes it easier to identify the key features of an image.

Bharti Sharma et.al. [7] proposed an algorithm for vehicle detection in which the filtering of the output image is done after applying the threshold value to the vehicle . Shape index is computed to extract the target object as it extracts the bright target object more precisely. Shape Index (SI) is calculated as,

$$SI = \frac{Perimeter}{\sqrt[4]{Area}}$$
(3)

By these operations vehicle is detected.

PushpalataPatilet.al. [8]explained a method for vehicle detection by image processing in which image enhancement is done with the help of a computer. The main objective is to create an image that is more suitable for specific applications. The technique implemented enhances the image by image's joints and lineaments. To emphasize brightness differences associated with linear features contrast enhancement is used.





LITERATURE SURVEY

BhavinkumarM.Rohit et.al [1] describes a system for detecting vehicles based on their rear lights. They used the segmentation approach and introduced low light video frames which have low exposure value images. The idea behind using the low exposure image frames is that factors such as street lights, unwanted reflections from vehicles body, sign boards reflections can be removed and only the bright red color and head lights of oncoming vehicles are visible in the image frame. Segmenting the red color from other noises is the major issue in identifying the tail lights. The red layer is firstly extracted from the rear light and is then converted to the gray image. The gray frame is then subtracted from the red frame and all the unwanted noises are removed by median filter. A threshold value is set to convert the obtained image to its corresponding binary image. Then the blob analysis techniques are used to calculate the area and the corresponding bounding boxes. Blob detection methods are aimed at detecting regions in an image that has similar as well as difference in their properties. Depending upon the symmetry, tail lights for same vehicle is identified and the nuisance are rejected.

Chen et al.[11] uses segmentation for identifying the bright object and verifies the segmented regions by spatial clustering based on the symmetric properties such as shape, texture and relative position. The Nakagami-m distribution approach is used to detect turn signals by scatter modeling of tail lights. The turn signals are detected using contrast enhancement in which intensity for the image is obtained. In order to avoid the noise generated from the non-tail lights a step function is applied and preprocessed. Thus obtained tail light is modeled using the Nakagami-m distribution. After this the color space regulation on the CIE xy chromaticity has been introduced to verify detected turn signals. For recognizing the direction of the detected turn signal as the reflectance strength of the area near turn signals is larger than that of other areas, vehicle reflectance is first decomposed. The bounding area is analyzed and in order to overcome the variation of an event pattern, a training algorithm is adopted like AdaBoost. Using this algorithm the classifiers of left and right turn signal directions are trained.

Noppakun Boonsim et.al. [12] present a new algorithm to detect a vehicle ahead by using taillight pair. Tail light detection is implemented by two steps tail light candidate extraction and tail light verification. The tail light candidate extraction phase includes extraction of color pixels from the input image and candidate extraction process segments red pixel regions containing white. By applying symmetry score of size, shape and position candidates the tail light verification is done. The symmetry of position is considered by estimating y-direction distance of each pair by,

DS
$$(P_k) = (1 - \frac{|H_i + H_j|}{|H_i + H_j|}) \times 100$$
 (4)

Where DS is y-distance symmetry score of pair P_k . H_i and H_j are the y axis-distance between border and center of C_i and C_j , respectively. Next, size symmetry is then checked to justify area characteristic equality using.

AS $(P_k) = (1 - \frac{|A_i - A_j|}{A_i + A_j}) \times 100$ (5)

(6)

Where AS is defined as size (area) symmetry score of pair $P_k A_i$ and A_j are the area of candidate C_i and C_j consecutively. Then the symmetry of shape is checked by analyzing aspect ratio of candidate width and height.

ARS
$$(P_k) = (1 - \frac{|ARS_i - ARS_j|}{ARS_i + ARS_j}) \times 100$$

Where ARS is the aspect ratio symmetry score of pair P_k , ARS_i and ARS_i represent the aspect ratio of candidate C_i and C_i respectively.

position symmetry score is experimentally defined as the most significant value following by size and shape symmetry Score

Symmetry Score
$$(SS)_k = 0.8 * DS_k + 0.1 * AS_k + 0.1 * ARS_k$$
 (7)

Finally, the maximum symmetry score will be considered to confirm the position of TLs. The symmetry score of TL should be more than 80. If the score is less than the threshold value, test images will be discarded. This is the equation to verify TL positions.

TL pair = Max (Symmetry Score
$$(P_k)$$
) > 80 (8)

Yen Lin Chen et.al.[13] explained an effective system for detecting and tracking moving vehicles during night has been explained. Using image segmentation and pattern analysis techniques the vehicles tail light are detected and identified. The main process done in this paper is the classification of car and motorbike tail lights.

A moving car is modeled as a rectangular patch, a tracked component group TG_k^t has been implemented and given by,

$$\tau_{r1} \le W \frac{(TG_k^t)}{H(TG_k^t)} \le \tau_{r2} \tag{9}$$

where τ_{r1} and τ_{r2} are threshold values 2.0, 8.0 and W, H represents the width and height of the bright objects. TG^t_k should be symmetrical and well aligned such that

$$\tau_{\alpha 1} \left(\frac{W(TG_k^t)}{H(TG_k^t)} \right) \le |TG_k^t| \le \tau_{\alpha 2} \left(\frac{W(TG_k^t)}{H(TG_k^t)} \right)$$
(10)

where $\tau_{\alpha 1}, \tau_{\alpha 2}$ are threshold values 0.4 and 2.0. The width of TG^t_k of a potential car should be in a reasonable ratio with respect to the lane width,

$$\tau_{\omega 1} LW(TG_k^t) \le W(TG_k^t) \le \tau_{\omega 2} LW(TG_k^t)$$
(11)

where $LW(TG_k^t)$ is the approximate lane width associated with TG_k^t . The threshold τ_{ω_1} and τ_{ω_2} are 0.5 and 0.9, respectively.

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To identify motorbikea single tracked component TP^t_i can be identified as,

$$\tau_{m1} \le \frac{W(TP_i^t)}{H(TP_i^t)} \le \tau_{m2} \tag{12}$$

where the threshold values τ_{m1} and τ_{m2} are 0.6 and 1.2.

Ronan O' Malley et.al.[14] explains segmenting red lamps and pairing to combine it with target vehicle. RGB color space is converted to HSV and cross relation method is used to pair the red light.

Table 1 Color thresholds

	Hue	Saturation	Value
Red	340°- 30°	0 - 30	80 - 100
White	All	0 - 20	99-100

Along the line adjoining the center region of light the cross relation is calculated. The cross relation matrix γ between two lamp image segments is calculated by,

$$\gamma = \sum_{x,y} \frac{(T(x,y) - \bar{T})(I(x,y) - \bar{I})}{\sigma_T \sigma_I}$$

(13)

CHALLENGES

Although many techniques have been implemented for vehicle detection during night, accidents caused by heavy vehicles can't be controlled yet. The methods implemented for detection are not used for heavy vehicles as their taillight orientation is different from that of cars and motorbikes. It is difficult to identify the vehicle if the lights are under repairing condition. The techniques available for detection uses computer vision techniques and so it requires high cost for its implementation.

CONCLUSION

This survey paper reviews different techniques used for the detection and identification of taillights of vehicles during night. Various vehicle detection methods are explained using the image segmentation techniques. In this paper many image processing techniques such as Segmentation, Edge detection, Filtering and image enhancement are discussed. By combining these methods the tail light are detected and the accidents during night are controlled up to some extent. More studies and researches are required in this field for further detection of all vehicles.

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