

# Lane Detection and Tracking using Recursive HOG Transform for Advance Driver Assistance



Sagar S. Tikar, Rajendra A. Patil

**Abstract:** Numerous individuals pass away every year in roadway crashes brought about by driver's absent mindedness. Path discovery frameworks are helpful in maintaining a strategic distance from these mishaps as wellbeing is the primary motivation behind these frameworks. Such frameworks have the objective to distinguish the path marks and to caution the driver on the off chance that the vehicle tends to leave from the path. A path location framework is a significant component of numerous smart vehicle frameworks. Path recognition is a difficult undertaking in light of the differing street conditions that one can run over while driving. In the previous barely any years, various methodologies for path discovery were proposed and effectively illustrated. Right now, a concise outline of existing strategies, we present a vigorous path discovery dependent on recursive HOG change. In path stamping acknowledgment, dimensional scale information, progressively changing area of plotting and recursive HOG change procedures are utilized to recognize path markings effectively. Trial results show that the proposed calculation is viable in picture pre-processing and can identify the path checking and vehicle precisely with less time.

**Keywords:** Recursive HOG transform; lane detection; lane departure; region of interest.

## I. INTRODUCTION

Presently a day the street mishaps have expanded all things considered. The greater part of the mishaps happen because of driver's carelessness and lack of regard while driving. Advance driver help framework (ADAS) assumes a significant job in giving security to drivers. It assists with robotizing the vehicle framework and builds the driving encounters. The Advance driver help framework (ADAS) gives a protected framework to decrease the street mishaps. The framework makes a functioning stride like admonition the driver or makes a restorative move to stay away from a mishap during the risky circumstance. The Lane Departure Warning (LDW) is a significant unit in Advance driver help framework. In vision based path flight framework, a camera is put behind the breeze shield of the vehicles and pictures of the street is caught. The white stripes out and about are deciphered and paths are distinguished. At whatever point the vehicle leaves the path then the admonition is given to the driver. In path takeoff notice framework, the path location is the underlying advance to be taken.

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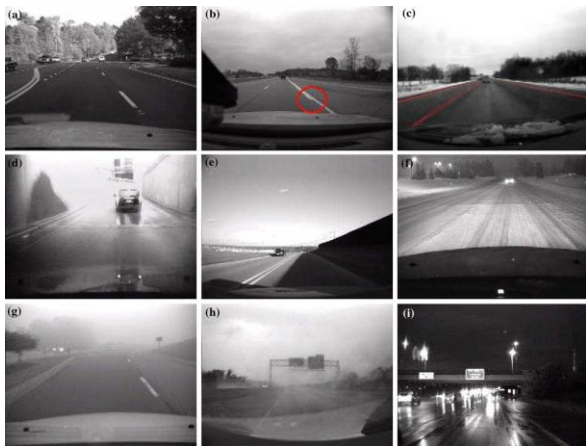
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There are two kinds of approaches utilized in path identification: the element based methodology and the model based methodology. The highlights based methodology distinguishes the path in the street pictures by recognizing the low level highlights, for example, path edges or painted paths and so on. This methodology requires all around painted lines or solid path edges, else it will fall flat. This methodology may experience the ill effects of impediment or commotion. The geometric parameters, for example, expecting the state of path can be introduced by straight line or bends are utilized by the model based drew closer.

## II. LITERATURE SURVEY

The speedy raise of urban traffic, the traffic security ends up being progressively imperative. Leaving the way causes about 30% of all disasters in the expressway, and an enormous part of these are come about on account of the interference and shortcoming of the driver. As needs be, a system that could give a reprimand to drivers of a hazard has a mind boggling potential to save a huge number of lives. Systems that are expected to help the driver in its driving technique are known as forefront driver help structures (ADAS). Various systems like flexible excursion control, crash avoidance structure, night vision powerless side area and traffic sign acknowledgment are a bit of ADAS [1]. Departure structure is also a bit of this characterization. This structure has a target to recognize the way checks and to provoke the driver if the vehicle will in general leave the way. Way acknowledgment is the system to discover way markers making the rounds and thereafter present these regions to a sharp structure. In insightful transportation structures [2], skilful vehicles help out wise establishment to achieve an increasingly secure condition and better traffic conditions. The uses of a way recognizing system could be as fundamental as pointing out way territories to the driver on an outside feature, to progressively complex assignments, for instance, envisioning a way change at the time future in order to avoid crashes with various vehicles. A part of the interfaces used to recognize ways join cameras, LASER go pictures, and LIDAR and GPS contraptions [3]. In many proposed systems, the way acknowledgment contains the constraint of explicit locals, for instance, road markings of the outside of the painted roads. Various challenges like left and moving vehicles, dreadful quality lines, shadows of trees, structures and various vehicles, more sharpened curves, flighty way shapes, mixing ways, works and various markings all over town, poor detectable quality conditions, for instance, turbulent condition, foggy condition, interesting black-top materials and dissimilar slopes destroys way ID.

There have been dynamic research on way revelation and a wide collection of counts of various depictions, distinguishing proof and following techniques, and modalities have been proposed [4].



**Figure (1): Some of the challenging scenarios for lane marking detection**

Vision-based way area techniques give insignificant exertion thickness information about avenues for free vehicles. At this moment, propose an energetic and beneficial method to develop the use of these procedures to cover low-speed circumstances. In any case, the strong zone near the vehicle is instated and a movement of rectangular revelation zones are intensely evolved along the road. By then, an improved adjusted neighborhood edge extraction is familiar with isolated the edge reasons for the way markings subject to correct stepping width limitations. In order to meet continuous requirements, a novel Bresenham line throwing a voting form space is proposed to improve the strategy of line parcel acknowledgment. Gotten together with straight lines, polylines, and twists, the proposed geometric fitting method can acclimate to various road shapes. Finally, one of a kind status vectors and Kalman channel move matrices are used to follow the key motivations behind the straight and nonlinear bits of the way. The proposed procedure was taken a stab at an open database and our self-overseeing stage. The exploratory results show that the methodology is incredible and beneficial and can meet the ceaseless necessities of independent vehicles. [5]

Yim and Oh [6] developed a three component based way area computation. The features used are starting position, heading and force regard. In the basic development, a Sobel director is applied to get the edge information. As far as possible is addressed as a vector including the three features. The current way vector is resolved reliant on the data picture and the past way model vector. Two windows, one for each, is used for left and right cutoff points. Tolerating  $N$  pixel in each even line,  $N$  way vector contenders are created. The best contender is picked subject to the base great ways from past way vector using a weighted detachment metric. For balance every segment is apportioned a substitute weight. By then a way acceptance system is used to predict the new way vector.

A way disclosure approach for urban condition is proposed by Sehestedt et al. [7]. Since the way markers are not unquestionably evident in view of mileage, hindrances and on account of complex road geometry, a fragile model is used for recognizing way markers. In the retrogressive perspective

mapped picture, particle channel is applied from base segment to top. The direct is tuned in such a way to follow various ways.

Aly [8] presents a continuous and incredible approach to bargain with recognize way markers in urban roads. It at first creates a top point of view all over town picture using reverse perspective mapping for keeping up a key good ways from the perspective effect. By then the top view is isolated using explicit masterminded two dimensional Gaussian piece. The channel is tuned unequivocally for splendid lines in diminish establishment with express width. So it has high response to the line markers and hold only the most raised characteristics by picking  $q$  % quantile regard from isolated picture and removing all characteristics underneath the edge. By then the straight lines are recognized using unraveled Hough change, which is trailed by RANSAC line fitting which gives early on speculation to the RANSAC spline fitting development. By then a post dealing with step is done to confine the spline and expand it in the image. The figuring doesn't perform following. It can recognize any number of way restrains in the image not just the current way.

Kim [9] developed a way acknowledgment and following computation which can manage testing circumstances, for instance, obscured way markers, way back and forth movements and separating ways. In the fundamental development, an incline discoverer and a force thump locator is used to discard the non way markers. Fake Neural Networks (ANN) is applied on extraordinary models for way acknowledgment. The recognized way markers pixels are assembled using cubic splines. Hypotheses are delivered from unpredictable course of action of line parts. RANSAC figuring helps in endorsing the theories. Particle isolating is used for way following.

Cheng et al. [10] introduced a different leveled computation for way area. High dimensional component centers are evacuated subordinate around feature concealing extraction. It is used to perceive composed lanes from unstructured lanes. By then related portions are applied on the component centers. A part vector is worked for feature centers outperforming an edge. Eigen regard Decomposition Regularized Discriminate Analysis is done to diminish the distinction of the model. By then most prominent likelihood Gaussian parameters are evaluated. The evacuated segment centers are used as recognized ways in sorted out avenues. For unstructured roads, the entire scene is parceled subject to mean-move division. Each region is seen as homogeneous and way markers are recognized using Bayes rule.

### III. PROPOSED SYSTEM

Essential point of this paper towards Lane location is under various light/climate conditions. Right now we examine various cases: the pictures are caught from the hybrid over the street, the paths to be identified can be straight or breath taking sharp curves, whenever for example day or night and with any climate conditions fortunate or unfortunate. The path markings can be strong or run lines.

- a. Image preprocessing and disappearing point extraction
- b. Area of intrigue extraction of the caught Image

- c. Lane checking discovery in caught pictures
- d. Lane location in video – reproduction by the model dependent on HOG change

Take a picture of street with the assistance of a camera fixed in the vehicle. Info picture is picture from database. Pre-preparing is a typical name for tasks with pictures at the most minimal degree of reflection both info and yield are force pictures. The point of pre-preparing is an improvement of the picture information that smothers undesirable expressions or upgrades some picture highlights significant for additional handling. Before talking about the extraction of highlight focuses it is important to have a measure to look at parts of pictures.

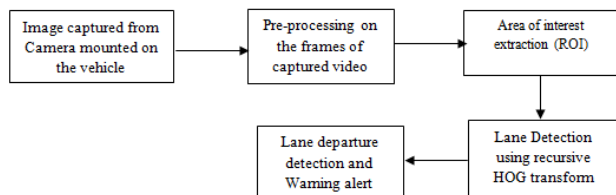


Figure (2): Architecture of proposed system

**Stage 1: Image pre-handling stage:** Here, we will probably clear disperse, deluding imaging objects and insignificant picture parts. The extra cleaned picture parts will be the data from which the fundamental features will later be isolated. All around, methodologies that fall under this current stage's degree can be requested into two distinct ways: dealing with light related effects on improve picture quality and cutting the bits of the image that are unessential for the further task. Another lighting up sway that ought not be dismissed is point of convergence flare which is realized by direct sunlight in the field of viewpoint on camera.

Various advances related with Image Pre-dealing with are:

- Enhance picture
- Weaken shadows
- Remove over and under introduction
- Remove deceiving picture objects
- Trimming the picture for ROI
- Remove superfluous areas.

**Stage 2: Region of Interest (ROI):** For example, lines on path are removed while the foundation is dismissed. Power of foundation influences working of framework, in night low force influences picture while high force in day time makes it difficult to perceive lines on street. For this Kalman channel is utilized to modify power.



(a) lane is present



(b) lane is absent



(c) lane present

Figure (3): Experimental results: lane detection of the image

Kalman channels are utilized to appraise states dependent on direct dynamical frameworks in state space design. The procedure model characterizes the development of the state from time  $k-1$  to time  $k$  as:

$$x_k = Fx_{k-1} + B\mu_{k-1} + w_{k-1} \quad (1)$$

where  $F_F$  is the state progress matrix applied to the past state vector  $x_{k-1}$ ,  $B$  is the control-input grid applied to the control vector  $\mu_{k-1}$ , and  $w_{k-1}$  is the procedure commotion vector that is thought to be zero-mean Gaussian with the covariance  $Q$

$$Q = w_{(k-1)} \sim N(0, Q) \quad (2)$$

Kalman channel calculation comprises of two phases: forecast and update. Note that the expressions "expectation" and "update" are frequently called "proliferation" and "remedy," individually, in various writing. The Kalman channel calculation is outlined as follows:

- Expectation:

Anticipated state estimate

$$x_k^A = Fx_{k-1}^A + B\mu_{k-1} \quad (3)$$

Anticipated blunder covariance

$$P_k^- = FP_{k-1}^+ F^T + Q \quad (4)$$

- Update:

Estimation residual

$$y_k^{\sim} = z_k - Hx_k^A \quad (5)$$

Kalman gain

$$K_k = P_k^- H^T (R + HP_k^- H^T)^{-1} \quad (6)$$

Refreshed state estimate

$$x_k^A+ = x_k^- + K_k y_k^{\sim} \quad (7)$$

Refreshed blunder covariance

In the above conditions, the cap administrator,  $\hat{\cdot}$ , implies a gauge of a variable.



That is,  $\hat{x}$  is a gauge of  $x$ . The superscripts – and ++ indicate anticipated (earlier) and refreshed (back) gauges, separately. Stage 3: HOG, or Histogram of Oriented Gradients, is a part descriptor, used to expel features from picture data. It is commonly used in PC vision assignments for object disclosure. The HOG descriptor revolves around the structure or the condition of an article. Crowd can give the edge course as well.

This is done by isolating the point and bearing of the edges. The HOG would make a Histogram for all of these regions autonomously. The histograms are made using the tendencies and headings of the pixel regards, thusly the name 'Histogram of Oriented Gradients'. The degree would be higher when there is a sharp change in power, for instance, around the edges.

$$\text{Total Gradient Magnitude} = \sqrt{[(G_x)^2 + (G_y)^2]} \tag{8}$$

Next, calculate the orientation (or direction) for the same pixel. We know that we can write the tan for the angles:

$$\tan(\phi) = G_y / G_x$$

Hence, the value of the angle would be:

$$\phi = \text{atan}(G_y / G_x) \tag{9}$$

So now, for each pixel esteem, we have the all out inclination (size) and the direction (heading). We have to produce the histogram utilizing these inclinations and directions. In the event that the direction of the ideal component is obscure, this strategy is muddle by the way that we should expand the aggregator by joining an additional parameter to represent changes in direction.



Figure (4): Determination of starting point of lanes

The assurance of the underlying paths is the key factor to the path identification. For the main edge, a far reaching approach joining HOG change and disappearing point is utilized to remove exact starting situation of paths. To start with, HOG change is applied to identify the lines, and afterward the convergences of these lines are determined. Be that as it may, it is difficult to get a remarkable crossing point moment that multiple lines exist. In this manner, streamlining methodology ought to be utilized, and the all out squared good ways from all the lines as a cost work is characterized in Equation (10),

$$I = \frac{1}{2} \sum (n_i^T (v_p - p_i))^2 \tag{10}$$

Where  $p_i$  signifies the point on line  $I$  recognized by HOG change and  $n_i$  is the unit ordinary to line  $I$ . At that point the evaporating point (VP) is chosen as the point whose cost work

is negligible. To locate the base, the cost work is separated as for the VP, so the declaration of disappearing point can be gotten with the accompanying condition.

$$v_p = (\sum n_i n_i^T)^{-1} (\sum n_i n_i^T P_i) \tag{11}$$

On account of complex enlightenment (extraordinary light, lopsided brightening, and so forth.) or genuine absence of the path lines, the HOG change will most likely be unable to identify all the lines of paths. As of now, estimation of the evaporating direct inclines toward fall flat, at that point the focal point of picture is taken as the disappearing point.

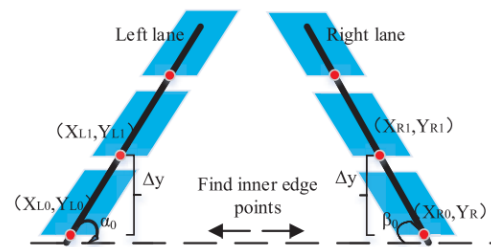


Figure (5): Illustration of AROI

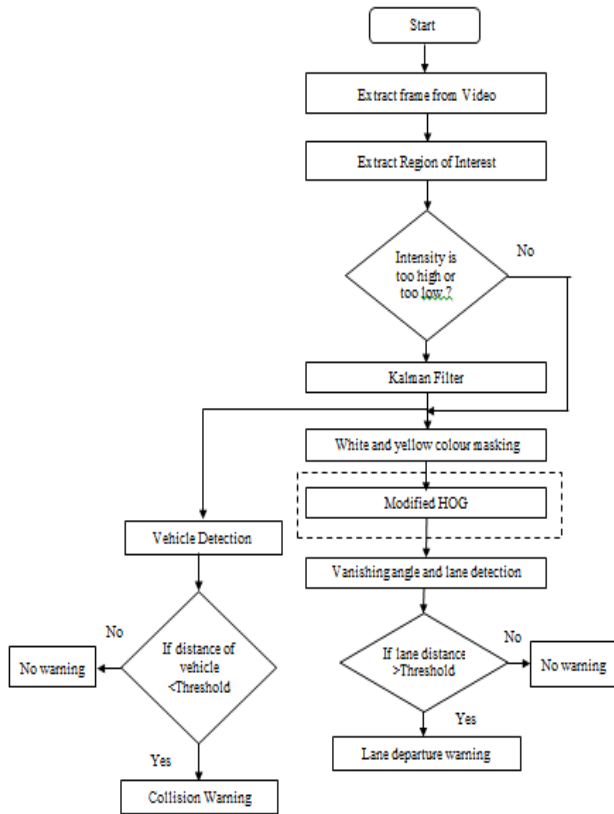
As showed by the information of the starting phase of way procured over, the primary ROI close to the host way can be settled, and a while later the motivations behind inside way edge are glanced in the ROI. By thusly, the uproar agitating impact can be discarded further. By then, we apply the least square strategy to fit these concentrations into a short line part. The purpose of this line segment, the circumstance of the past ROI, and the between time  $1y$  can be used to choose the accompanying ROI, as showed up in Equation (12), which acknowledges the left way for example. This allows a straight desire for the ROI along the course of the way, as portrayed in Fig. 5.

$$\begin{cases} X_{L1} = X_{L0} + \frac{\Delta y}{\tan \alpha} \\ Y_{L1} = Y_{L0} - \Delta y \end{cases} \tag{12}$$

To achieve the versatile after of the straight or twist way, the discrete AROIs are introduced, whose size and position can be adjusted continuously according to the stream change of way and the speed of the vehicle. Test outcomes infer that the shape change of the straight way is close to nothing, so we a little bit at a time decrease the width of ROI and suitably increase the interval between ROIs. While the rhythmic movement of the twist way moves phenomenally, we increase the width of the ROI and abatement the break between ROIs to keep away from missing area.

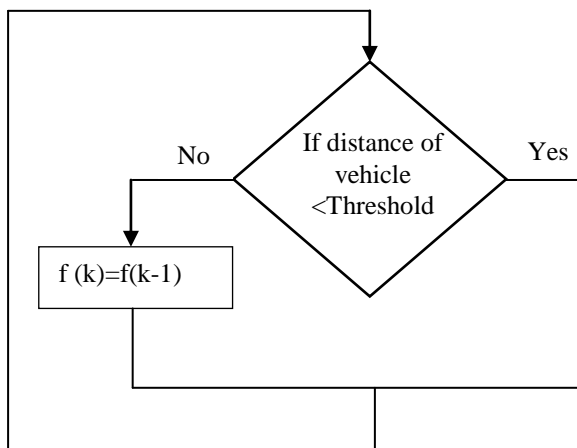
After the ways are removed, a limited critical level depiction of the way can be obtained, which is used for dynamic. The edges of way lines procured from each rous are utilized to perceive the way as straight or curve. Exactly when the distinction in edge is more noticeable than a particular worth ( $10^\circ$  in our investigations), the least square twist model is gotten to fit the way. Something different, the least squares direct model is gotten.

The intelligent activities at each progression are explained in figure (6) beneath:



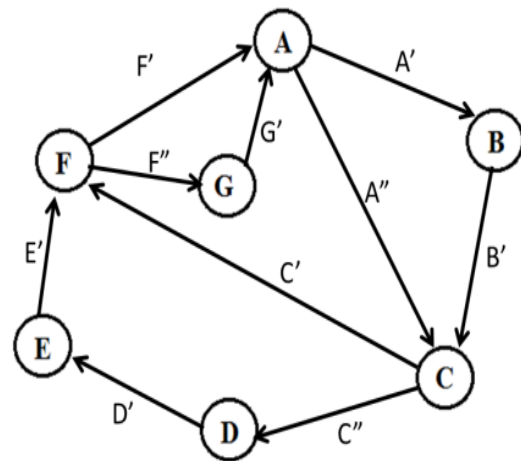
**Figure (6): logical operation of detection and warning method for lane and vehicle**

Modified HOG transform is applied on curves or lines drawn on lane. If succeeding line is vanished or not drawn, in that case we are considering previous line is continuing and recursively (repeatedly) use its equation and values unless new line isn't seen as illustrated in figure below. Hence recursive HOG transforms is used.



**Figure (7): Example of recursive HOG transform State Diagram:**

Input video stream is first check, if the intensity is proper, state C will be the next state after state A otherwise video first processed through state B and then to state C. if lane is detected then state F will be the following state otherwise statistical parameters of previous lane (calculated in state D) are used to predict next lane (state E). if travelling is not within lane then the warning signal will be given (state G).

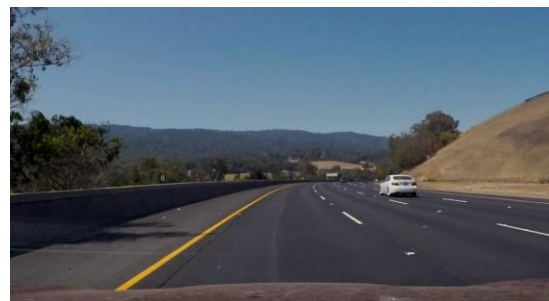


**Figure (8): State diagram of proposed system**

The states indication and meaning of every state is given below:

- A=input video stream, B=Kalman filter
- C=lane detection, D=get statistical data of previous lane
- E=predict lane using recursive HOG, F=calculate lane distance threshold
- G=lane departure warning, A'=intensity too low or too high
- A''=proper intensity, B'=modified intensity level stream
- C'=detected lane, C''= lane doesn't detected
- D'= statistical parameters of previous lane
- E'= predicted lane, F'=travelling within lane
- F''=travelling is not in limited lane range
- G'=warning signal

**IV. EXPERIMENTAL RESULTS AND ANALYSIS:**



**Result (1): Frame with light shadows and 'No warning' for lane departure**



**Result (2): Frame with shadows and 'No warning' for lane departure**



**Result (3): Frame with darker shadows and 'Warning' for lane departure**



**Result (4): Frame with missing marking and 'Warning' for lane departure**

Result (1) to (4) has demonstrated the output of the enhanced model of lane detection using recursive HOG method. The results are quite better than the conventional methods.

## V. CONCLUSION:

Lane departure cautioning is an unpreventable module in the impelled driver help systems. In the latest decade a couple of types of progress occurred in the way disclosure and following field. Vision based approach is an essential system for perceiving ways. In spite of the way that piece of progress has been cultivated in the way disclosure and following region, there is still augmentation for development due to the wide extent of variability in the way conditions. Vanishing point is resolved using trapezoidal rule, if dissipating point and focal point of huge side of trapezoid are on same line then driver is going right way. Kalman channel is used to manage power properly. The force out of the blue changes when whether changes phenomenally in stormy season, to avoid this sudden change here we use HOG change.

At the present time, practical system using the piecewise straight expanding limit with regards to way disclosure is presented. Hough change is used to perceive the way markings.

- The system was inspected under various conditions of developing edification, and shadows impacts in various road types.
- The system has demonstrated a healthy introduction for perceiving the road ways under different conditions.
- The time and accuracy for way disclosure and vehicle ID methodology is improved by division of zone of interest strategy.
- A game plan of investigation showed that the ways were perceived using HOG change. From the above result, we find the count capacities honorably for all lighting conditions like poor lighting up, with shadows, too much marvelous and with way missing too. The preliminary outcomes exhibited that the structure can achieve a standard need to give significant information to the driver to ensure security.

## REFERENCES:

1. F. Mariut, C. Fosalau and D. Petrisor, "Lane Mark Detection Using Hough Transform", In IEEE International Conference and Exposition on Electrical and Power Engineering, pp. 871 - 875, 2012.
2. S. Srivastava, R. Singal and M. Lumb, "Efficient Lane Detection Algorithm using Different Filtering Techniques", International Journal of Computer Applications, vol. 88, no.3, pp. 975-8887, 2014.
3. Borkar A, M. Hayes, M.T. Smith and S. Pankanti , "A Layered Approach To Robust Lane Detection At Night" , In IEEE International Conference and Exposition on Electrical and Power Engineering, Iasi, Romania, pp. 735 - 739, 2011.
4. Z. Kim, "Robust Lane Detection and Tracking in Challenging Scenarios", In IEEE Transactions on Intelligent Transportation Systems, vol. 9, no. 1, pp. 16 - 26, 2008.
5. Qingquan Li, Jian Zhou, Bijun Li, Yuan Guo and Jinsheng Xiao Robust Lane-Detection Method for Low-Speed Environments Sensors 2018, 18, 4274; doi:10.3390/s18124274 [www.mdpi.com/journal/sensors](http://www.mdpi.com/journal/sensors)
6. Y.U. Yim and S.- Y. Oh, "Three-feature based automatic lane detection algorithm (TFALDA) for autonomous driving," IEEE Trans. Intell. Transp. Syst., vol. 4, no. 4, pp. 219-225, Dec. 2003.
7. S. Sehestedt, S. Kodagoda, A. Alempijevic, and G. Dissanayake, "Robust lane detection in urban environments," in Proc. IEEE Intell. Robots Syst., Oct. 2007, pp. 123-128.



8. M. Aly, "Real time detection of lane markers in urban streets," in Proc. IEEE Intell. Veh. Symp., Jun. 4–6, 2008, pp. 7-12.
9. Z. Kim, "Robust lane detection and tracking in challenging scenarios," IEEE Trans. Intell. Transp. Syst., vol. 9, no. 1, pp. 16-26, Mar. 2008.
10. H.Y. Cheng, C.C. Yu, C. C. Tseng, K. C. Fan, J. N. Hwang, and B. S. Jeng, "Hierarchical lane detection for different types of roads," In Acoustics, Speech and Signal Processing (ICASSP), 2008 IEEE International Conference on pp. 1349-1352. IEEE, 2008.
11. Umar Ozgunalp, Rui Fan, Xiao Ai, and Naim Dahnoun "Multiple Lane Detection Algorithm Based on Novel Dense Vanishing Point Estimation", IEEE TRANSACTIONS ON INTELLIGENT TRANSPORTATION SYSTEMS, VOL. 18, NO. 3, MARCH 2017 621
12. M. Meuter, S. Muller, A. Mika, S. Hold, C. Nunn and A. Kummert, "A Novel Approach to Lane Detection and Tracking", In IEEE 12th International Conference on Intelligent Transportation Systems, pp. 1-6, 2009.
13. S. Zhou, Y. Ziang, J. Xi, J. Gong, G. Xiong and H. Chen, "A novel lane detection based on geometrical model and gabor filter", in IEEE Intelligent Vehicles Symposium, pp. 59-64, 2010. [19] Z. Teng, J.H. Kin and D.J. Kang, "Real-time Lane detection by using multiple cues", In IEEE International Conference on Control Automation and Systems, pp. 2334 - 2337, 2010.
14. N. Phaneendra, G. Goud and V.Padmaja, "Accident Avoiding System Using Lane Detection", International Journal of Research in Electronics and Communication Engineering, vol. 1, no. 1, pp. 1 - 4, 2013. [21] D. Pomerleau and Jochem, "Rapidly Adapting Machine Vision for Automated Vehicle Steering, IEEE, 1996.
15. B. M. Broggi, "GOLD: A parallel real-time stereo Vision system for generic obstacle and lane detection", IEEE Transactions on Image Processing, pp. 4-6, 1998.
16. C. Kreucher and S. K. Lakshmanan, A Driver warning System based on the LOIS Lane detection Algorithm, Proceeding of IEEE International Conference On Intelligent Vehicles. pp. 17 -22, 1998.
17. Y. Wang, E. K. Teoha, D. Shen. Lane detection and tracking using B-Snake", In: Image and Vision Computing 22, pp: 269-28, 2004. M. Chen., T. Jochem and D. T. Pomerleau, "AURORA: A Vision-Based Roadway Departure Warning System", In Proceeding of IEEE Conference on Intelligent Robots and Systems, 2004. [26] C. R. Jung and C. R. Kelber, "Lane following and lane departure using a linear-parabolic model" In: Image and Vision Computing, pp. 1192-1202, 2005.

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