

A Deep Learning Application for Wrong Route Vehicle Detection Using Hough Transform and Lucas Kanade Algorithms

Mr. B. Ajayram(M.Tech)¹, G.S.Karthik²,G.V.U.Mahesh³,
D.Bhavya⁴, G.L.K.Mohan⁵

Assistant Professor¹, LENDI INSTITUTE OF ENGINEERING AND TECHNOLOGY,A.P.,India.
Student^{2,3,4,5},B.Tech(CSE),LENDI INSTITUTE OF ENGINEERING AND TECHNOLOGY.

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ABSTRACT—

Studies have shown that installing wrong-way vehicle detection systems can help turn more than 67% of drivers around before they enter the freeway. The software can easily detect any vehicle if that particular vehicle traveling in the wrong direction. If the rule is not followed properly, an alarm will be generated at the nearest junction. If the particular traveller travels in the wrong direction beyond the limit, then the system detects by using a vehicle number and sends a message(E-mail) to that particular vehicular. All these things are happening by taking a recorded video, we can extract frames from that video and find the traffic divider line position using Hough Transform and then determine the mask of left and right sides of the road using the traffic divider line and apply the Lucas Kanade Optical Flow to determine whether the direction of the vehicle is correct or not. If any traveller travels in the wrong direction beyond the limit, at that time vehicle number is extracted by using the OCR technique with that we can get the vehicular info and furtherly we perform an action on them.

KEYWORDS— Hough Transform ,Lucas Kanade, Optimal Recognition Character.

I. INTRODUCTION

In these days Road accident is a very common issue in a densely populated country. In 2020, there were 3702 roadaccidents leading to 4227 deaths including many children and students. The capacity of the roadway is not sufficient for the growing number of vehicles and thus imbalance is created. The drivers do not follow traffic rules and take advantage of driving in the wrong side in times of red traffic signals. It increases traffic on one side and hampers traffic flow greatly. It also increases the possibility of head-on collision several times. About 355 people die every year due to crashes in wrong-way driving in the United States. This is a common scenario in most of the cities. So, it's essential to prevent drivers from driving on the incorrect side. To ensure it, those who don't follow traffic rules need to find out and strict law should be applied.



Fig: Heavy traffic jam due to wrong-way driving

The system we present here has following stages. In the first stage, every vehicle in the video frame is detected using the Hough Transform and Lucas Kanade algorithms. Alarm will be generated at nearest junction if that particular vehicle detects (i.e if the vehicle travels in a wrong direction). The direction of the vehicle is determined by using Hough Transform and detect whether it moves in the wrong direction or not. If the vehicle is on the wrong side, then the system will capture an image of the vehicle. And also alert message is send to that particular vehicle member. So, this creates a less chances for accidents.

II. RELATED WORK

Deep Learning technology is being used in many intelligent traffic monitoring systems. So far, many systems have been developed for wrong-way vehicle detection. The existing methods can be classified into sensor-based and motion pattern-based. Sensor-based methods use a magnetic sensor to detect the direction of vehicles. The earth's magnetic field gets disturbed within the presence of a vehicle and therefore the sensor gives the signal according to the variation of the magnetic field. From the signal, the direction of the vehicle is often detected. But in this system, the magnetic field can be varied for other reasons too. In that case, it won't give a satisfactory result. The motion pattern-based method uses optical flow measurement. The method proposed uses optical flow calculation to detect the direction and compare it with the modeled lane direction. But this method faces a drag thanks to occlusion. enhance optical flow estimation was used and the authors used background subtraction and the Lucas-Kanade method to detect the wrong-way vehicles.

III. EXISTING SYSTEM AND LIMITATIONS

According to the existing process, it is just going to detect a vehicle which is going in Wrong Route. It also alerts with an alarm sound when a vehicle is detected in a wrong route. But it is unable to store the records of the vehicles which were into the wrong route in the past and it didnot send warning message and fine details to the vehicular.

IV. PROPOSED SYSTEM

In existing process, it is just going to detect a vehicle which is going in Wrong Route. It also alerts with an alarm sound when a vehicle is detected in a wrong route. But it is unable to store the records.

In our project along with detecting the vehicle that goes in the wrong route, we are going to save the data records of the particular vehicles. Also, we are going to save the traveller /Owner information of the detected vehicles. And also we will send an alert to the Vehicular if he drives in a wrongroute.

1. Selecting the target area:

At first we choose the target area in order to record the video. The region should have a reasonable distance if not there will be a problem to track the large vehicles.

2. Extract frames from a video frame:

By using that video, we extract frames from a input video. In our project we extract frames by using OpenCV module in python.

3. Performing edge smoothing and edge detection:

In order to perform Hough transform we should know the lines or edges which exists on an image. To remove a smoothing sharp noise from that image we do edge smoothing before finding the edges. Edge detection was performed by finding the x and y gradient of an image. The threshold after edge detection to remove unnecessary small edge like the edges on a car or a road which isnot useful for Hough transform.

4. Implementation of Hough Transform Algorithm:

Now, perform hough transform on the edge detected image. Then we find the traffic divider line position using Hough transform. And choose 2 frame analysis from extracted frames to find the masks of left and right side of the road using that traffic divider line. Before applying optical flow method, perform Gaussian smoothing on each image because smoothing the images removed any sharp noises and intensities and we get smoothed derivatives finally.

5. Apply Lucas Kanade Algorithm:

Now, Apply Lucas Kanade algorithm (least squares) in order to detect the vehicle i.e if any vehicle travels in a wrong direction.

Step 1: Apply median filter on u and v

Step 2: Choose the degree with maximum count. If maximum degree less than 0 vehicle is in right direction else vehicle is in wrong direction.

6. Extraction of Vehicle Number:

If the vehicle travels in the wrong direction we detect the vehicle and then we capture the vehicle number. Now, by using that we get vehicle records and we send a message to that particular traveller (message consists of fine with a due date).

V. ADVANTAGES OF PROPOSED SYSTEM

- It automatically sends an alert message about a vehicle that crossed beyond limits to a nearby police station.

- It automatically generates message regarding fine to the vehicle owner.
- With this atleast few people will follow the rules strictly.
- With this there will be downfall in accident graph definitely

VI. HOUGH TRANSFORM & LUCAS KANADE

The Hough Transform is an algorithm patented by Paul V. C. Hough and was originally invented to acknowledge complex lines in photographs (Hough, 1962). Since its inception, the algorithm has been modified and enhanced to be able to recognize other shapes such as circles and quadrilaterals of specific types. So as to know how the Hough Transform algorithm works, it's

important to know four concepts: edge image, the Hough Space and therefore the mapping of edge points onto the Hough Space, an alternate thanks to represent a line, and the way lines are detected. The Hough Space could be a 2D plane that contains a horizontal axis representing the slope and therefore the vertical axis representing the intercept of a line on the original image. The edge point produces a line in Hough Space within the kind of $b = ax_i + y_i$ (Leavers, 1992) within the Hough Transform algorithm, the Hough Space is employed to see whether a line exists within the edge image. The Hough transform may be a feature extraction technique utilized in image analysis, computer vision, and digital image processing. Using this algorithm, we find the traffic divider line position in the extracted video frame.

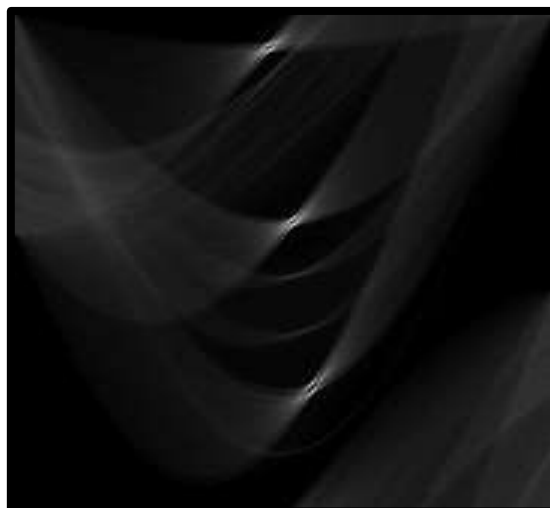


Figure displays Hough Space generated from one of the selected frames

In computer vision, the Lucas-Kanade method could be a widely used differential method for optical flow estimation developed by Bruce D. Lucas and Takeo Kanade. It assumes that the flow is actually constant in a very local neighbourhood of the pixel under consideration, and solves the fundamental optical flow equations for all the pixels therein neighbourhood, by the smallest amount squares criterion. By combining information from several nearby pixels, the Lucas-Kanade method can often resolve the inherent ambiguity of the optical flow equation. It's also less sensitive to image noise than point-wise methods. On the opposite hand, since it's a purely local method, it cannot provide flow information within the interior of uniform regions of the image. It works by trying to guess during which direction an object has moved so that local changes in intensity are often explained. Assume that we

watch a scene through a square hole. The intensity a visual through the outlet is variable. Within the next frame the intensity of the pixel has increased to b . It's the instantaneous velocity of the pixel movement of the spatially moving object on the observation imaging plane. It uses the change of the pixel within the time domain within the image sequence and therefore the correlation between adjacent frames to search out the existence between the previous frame and therefore the current frame correspondence, so on calculate the motion information of objects between adjacent frames. Generally speaking, optical flow is caused by the movement of the foreground object itself, the movement of the camera, or the joint movement of the 2 within the scene. The optical flow algorithm evaluates the deformation between two images, and its supposal is that the conservation of voxels and image pixels. It assumes that the colour of an object

doesn't change significantly and significantly within the previous two frames. supported this concept, we will get the image constraint equation. Different optical flow algorithms solve the optical flow problem assuming different additional conditions. In 1981, Horn and Schunck creatively linked the two-dimensional velocity field with the grey level, introduced the optical flow constraint equation, and obtained the fundamental algorithm for optical flow calculation. Assuming that the pixel intensity of the target within the image doesn't change between consecutive frames and let $I(x,y,t)$ denote the grey value of the pixel at time t , The assumption of constant brightness is on no account true in point of fact, but it is more reasonable and natural. As long as Δt is little enough, this assumption can basically be met. We perform a first-order Taylor expansion. Since this equation has two unknowns, there is no unique solution. so as to urge unique solution, new constraints or assumptions must be added. This

algorithm effectively finds the sides of the image which contain the foremost information. The Lucas Kanade implementation could keep track of targets throughout affine transformations and occlusions. Although there are still some improvements to be made to handle occlusion and maintain stability. Lucas Kanade tracker when the tracker is not any longer able to maintain an honest track on the target. The Lucas Kanade tracker doesn't require a particle filter. The first mode will take the ultimate output of the Lucas Kanade tracker and use those as inputs to the particle system measurements. The subsequent frame predicts the constant velocity motion model. Weight each particle supported its distance from the Lucas Kanade targeting output Using the current frame. The Gaussian random distribution for speed base. Resample the particles based on weight. Lucas Kanade algorithm is used to determine whether the direction of the vehicle is correct or not.

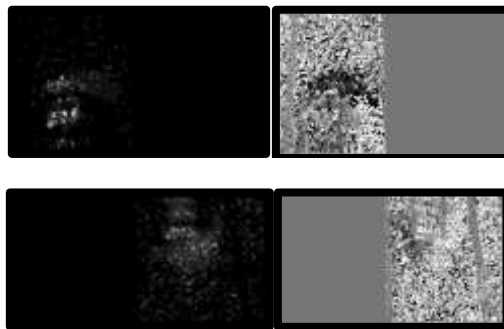


Figure showing applying median filter on u and v to the left side and right side of the road.

OCR, or Optical Character Recognition, could be a process of recognizing text inside images and converting it into an electronic form. These images may well be of handwritten text, printed text like documents, receipts, name cards, etc., or perhaps a natural scene photograph. OCR has applications in a very broad range of industries and functions. So, everything

from scanning documents – bank statements, receipts, handwritten documents, coupons, etc., to reading street signs in autonomous vehicles – this all falls under the OCR umbrella. OCR is the technology which extracts alpha numeric characters for vehicle number plate visible in images/videos & live camera streams too.

VII. RESULT

CASE-1:



The figure shows vehicle on the left side of the road is in correct direction and vehicle on the right side of the road is in correct direction.

CASE-2:



The figure shows vehicle on the left side of the road is in wrong direction and vehicle on the right side of the road is in correct direction.

VIII. CONCLUSION

As our country is one of the most populated country in which the deaths rate due to accidents high. So we are came up with an idea to reduce the accidents by implementing this project. This is about identifying the vehicles going in wrong direction gives an alarm sound. It generates an automated fine message for vehicle owner which are moving in wrong direction. The two road areas of each frame were sequentially detected to obtain good vehicle detection results in the monitoring field. The position of the object in the image was predicted by the Lucas Kanade algorithm based on the object detection result. The results verified that the proposed vehicle detection and tracking method for highway surveillance video scenes has good performance and practicability. Compared with the traditional method of monitoring vehicle traffic by hardware, the method of this paper is low in cost and high in stability and does not require large-scale

construction or installation work on existing monitoring equipment.

REFERENCES

- [1]. Z. Rahman, A. M. Ami, M. A. Ullah, "A Real-Time Wrong-Way Vehicle Detection Based on YOLO and Centroid Tracking" published IEEE Region 10 Symposium (TENSYP)
- [2]. S. Usman, S. Baydadaev, "Real-Time, Deep Learning Based Wrong Direction Detection" published in Applied sciences 2020
- [3]. L. W. Tsai, J. W. Hsieh, and K. C. Fan, "Vehicle detection using normalized color and edge map," IEEE Transactions on Image Processing, vol. 16, no. 3, pp. 850–864, 2007.
- [4]. X. Wen, L. Shao, W. Fang, and Y. Xue, "Efficient feature selection and classification for vehicle detection," IEEE Transactions on Circuits and Systems for Video Technology, vol. 25, no. 3, pp. 508–517, 2015.
- [5]. C. Papageorgiou and T. Poggio, "A trainable system for object detection," International Journal of Computer Vision, vol. 38, no. 1, pp. 15–33, 2000.