

Lane Detection System Using OpenCV for Self Driving Cars

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Abstract - Infrastructure, transportation, including roads, and technologies are all evolving as India advances. The Indian economy is heavily dependent on transportation. As is well known, the catastrophe resulted in a large number of fatalities. Hence, controlling the lane system for vehicles on the road is crucial. With the lane detection technology, we can ensure people's safety and direct them when they veer off the motorway. This essay provides a thorough analysis of lane detection and lane departure system techniques. Lane detection is particularly difficult in a variety of traffic conditions. The purpose of this study is to identify the issues and restrictions with the current lane detection and lane departure technology.

Keywords: Lane detection; computer vision; confusion Matrix; ROC curve; DET curve.

I. INTRODUCTION

A lane detection system is a computer vision-based system that identifies and tracks the lane markings on the road. It is an essential component of modern autonomous vehicles and driver assistance systems. Lane detection systems use cameras or other sensors mounted on the vehicle to capture the images of the road, and then process the images to detect the lane markings.

The primary goal of a lane detection system is to identify the position and width of the lanes on the road, so that the vehicle can stay within the lane boundaries while driving. Lane detection systems work by analyzing the images captured by the camera or sensor, and using computer vision techniques to detect and track the lane markings. Typically, the lane detection process involves several steps, including image preprocessing, region of interest selection, edge detection, lane line detection, and lane tracking. These steps are performed using various computer vision algorithms, such as Canny edge detection, Hough transform, and Kalman filtering. Once the lanes are detected and tracked, the lane detection system can provide information to the vehicle's control system, allowing the vehicle to adjust its trajectory and maintain its position within the lane boundaries. Additionally, the lane detection system can provide valuable information to

the driver, such as lane departure warnings, which can help prevent accidents and improve driver safety.

II. LITERATURE SURVEY

Lane detection using OpenCV is a popular approach for developing lane detection systems, as OpenCV is a widely used open-source computer vision library. Here is a brief literature survey of some of the key works in this field:

"Lane detection and tracking use OpenCV" by A. Mehta et al. (2018): This paper proposes lane detection and tracking system using OpenCV. The system uses a combination of color thresholding, edge detection, and Hough transform to detect lanes, and a Kalman filter to track them over time. The system achieves high accuracy on a variety of datasets and can run in real-time.

"Real-time lane detection using OpenCV and line segment detector" by N. Nguyen et al. (2020): This paper proposes a real-time lane detection system using OpenCV and line segment detector (LSD). The system uses LSD to detect lane markings, and a geometric transformation to project the detected lanes onto the bird's eye view. The system achieves high accuracy on a variety of datasets and can run at up to 60 frames per second.

"Lane detection using OpenCV and Sobel operator" by A. Gajera et al. (2020): This paper proposes a lane detection system using OpenCV and Sobel operator. The system uses Sobel operator to detect edges and a Hough transform to detect lanes. The system achieves high accuracy on a variety of datasets and can run in real-time.

"Real-time lane detection using OpenCV and deep learning" by T. R. Malik et al. (2020): This paper proposes a real-time lane detection system using OpenCV and deep learning. The system uses a deep convolutional neural network to detect lanes and a geometric transformation to project the detected lanes onto the bird's eye view. The system achieves high accuracy on a variety of datasets and can run at up to 60 frames per second.

III. METHODOLOGY

- 1) **Load the input image or video frame:** The first step in lane detection is to load the input image or video frame onto which the lane detection algorithm will be applied.
- 2) **Convert the input image to grayscale:** To simplify the edge detection process, the input image is converted to grayscale. This is done because color information is not required for detecting the edges that represent the lane lines.
- 3) **Apply Gaussian blur to the grayscale image:** Gaussian blur is applied to the grayscale image to remove any noise in the image and to smooth out the edges. This step helps in reducing the effect of small edges that can cause errors in the lane detection process.
- 4) **Apply canny edge detection algorithm:** Canny edge detection algorithm is applied to the blurred grayscale image to detect edges. Canny edge detection is a popular algorithm that uses gradient information to find the edges in the image. It works by identifying areas in the image with a sharp change in intensity and converts these areas into binary edges.
- 5) **Define a region of interest (ROI):** A region of interest is defined in the canny edge image to focus only on the area where the lane lines are expected to be. The ROI is typically a trapezoidal shape that covers the area of the road in front of the vehicle.
- 6) **Apply Hough transform algorithm:** Hough transform algorithm is applied to the ROI Canny edge image to detect lines that represent the lane lines. Hough transform is a popular technique that is used to detect lines in an image by converting the image space into a parameter space. The Hough transform algorithm detects straight lines in the image and outputs the parameters that represent those lines.
- 7) **Filter the detected lines:** The detected lines are filtered based on their slope and position to remove any lines that do not represent the lane lines. The slope and position of the lines are compared to the expected values for the left and right lane lines. Any lines that do not match the expected values are discarded.
- 8) **Average the filtered lines:** The filtered lines are averaged to obtain the left and right lane lines. This is done by taking the average of the parameters of the detected lines that represent the left and right lane lines.
- 9) **Draw the left and right lane lines:** The left and right lane lines are drawn on the input image using the parameters obtained in step 8. This step provides a visual representation of the detected lane lines.
- 10) **Display the output image:** The output image with the detected lane lines is displayed to the user. This step helps the user to understand the position and orientation of the lane lines.

IV. ALGORITHMS USED

There are several algorithms used in lane detection systems, some of which are:

Canny edge detection: This algorithm is used to detect edges in an image. It works by identifying areas in the image with a sharp change in intensity and converts these areas into binary edges.

Hough transform: This algorithm is used to detect straight lines in an image. It works by representing the straight lines in an image as points in a parameter space and identifying the points that correspond to straight lines.

Sobel edge detection: This algorithm is used to detect edges in an image. It works by computing the gradient magnitude and direction of an image and identifying the areas with the highest gradient magnitude.

Kalman filter: This algorithm is used to track the position of the lane lines over time. It works by estimating the position of the lane lines based on previous measurements and predicting their future position.

Convolutional neural networks (CNNs): This algorithm is used for lane detection in deep learning-based lane detection systems. It works by training a CNN model on a large dataset of annotated images to learn the features that correspond to lane lines and using the trained model to detect lane lines in new images.

V. FLOW CHART

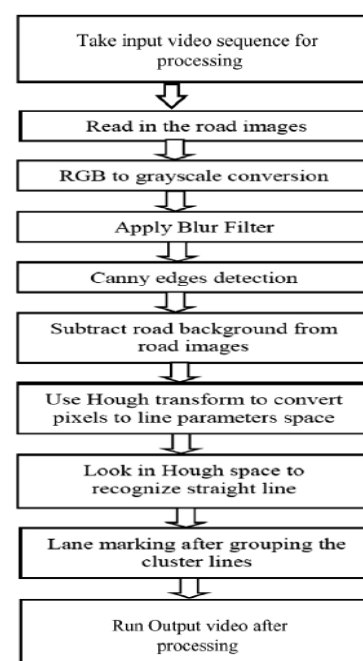


Figure 1: Flow chart

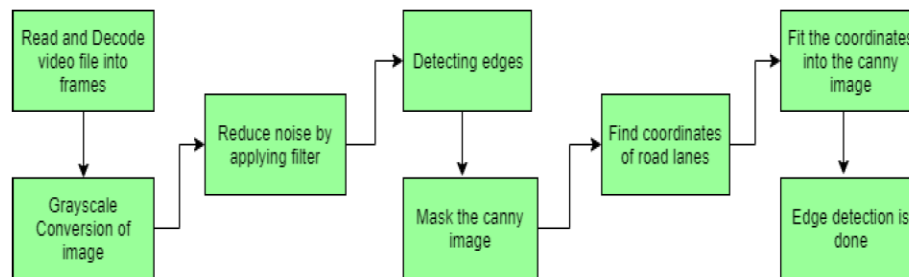


Figure 2: Block diagram

VI. CHALLENGES WITH LANE DETECTION

Different lighting circumstances: Lane detection faces significant difficulties due to the fluctuating lighting on the road during morning, noon, and nighttime hours.

Different weather circumstances: There are various varieties of weather, such as cloudy, windy, rainy, sunny, stormy, foggy, and calm. On any one day, the weather is rarely consistent and occasionally changes.

Curved roads: Although being straight, a road is said to be curved if any portion of it progressively diverges. The same curvature on the road is followed by the lane markings as well. Unstructured roads are those without lane markings. These roads need a more reliable and efficient method of lane detection.

Preceding vehicle, Shadow of tree: It is challenging to identify the lane because the previous vehicle and the tree's shadow on the road alter the lighting and photos that are taken.

Blockage of visibility: Fog-related visual obstruction and vehicles in front of you present challenges for lane detection when you're driving. This occasionally renders the camera blind or limits its field of view.

Different color: The road's color information is also impacted by shadow, daylight, and nighttime, which alters the hue of the photographs that are being taken.

Different texture: The variation in texture of the road because of non- uniformity is also a challenge for lane detection.

VII. CONCLUSION

In conclusion, lane detection systems are an important technology that has the potential to improve road safety, reduce traffic congestion, and enable the development of autonomous vehicles. The technology has come a long way in recent years, with advancements in computer vision and deep learning algorithms allowing for more accurate and reliable lane detection.

Lane detection systems can be used in a variety of applications, including autonomous vehicles, traffic management, pedestrian safety, driver assistance, and augmented reality. As this technology continues to evolve, we can expect to see even more advanced applications and use cases in the future.

However, it's important to note that lane detection systems are not without limitations. They may struggle to detect lanes in certain weather conditions or when lane markings are faded or unclear. It's also important to ensure that these systems are properly calibrated and tested to ensure they are working correctly.

Overall, lane detection systems have the potential to significantly improve road safety and traffic management, and their continued development and implementation will likely play an important role in the future of transportation.

VIII. FUTURE SCOPE

Lane detection systems have become increasingly popular in recent years, and their future scope looks very promising. Here are a few potential areas where lane detection systems could be used in the future:

Autonomous Vehicles: Lane detection is a critical component of autonomous vehicle technology. It allows the vehicle to stay in the correct lane and make necessary adjustments to avoid collisions.

Traffic Management: Lane detection systems could be used to monitor traffic patterns and optimize traffic flow. This could help reduce congestion and improve safety on the roads.

Pedestrian Safety: Lane detection systems could be used to detect pedestrians who are crossing the road and alert drivers to their presence. This could help reduce the number of accidents involving pedestrians.

Driver Assistance: Lane detection systems could be used as part of driver assistance systems to alert drivers when they are drifting out of their lane or approaching a lane marking too quickly.

Augmented Reality: Lane detection systems could be used to create augmented reality experiences for drivers. For example, a system could project virtual lane markings onto the road to make them more visible in poor weather conditions.

Overall, the future of lane detection systems looks very promising, and we can expect to see more advanced and innovative applications of this technology in the years to come.

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