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# **Intelligent Highway Watch Project**

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**Course: DS501/AIML Lab**



Introduction

## Project Overview

Our project addresses the critical challenge of automated **traffic violation detection during nighttime conditions**. By combining computer vision and machine learning techniques, we've developed a system that can effectively monitor traffic signals, detect stop line violations, and identify violating vehicles through **license plate recognition**.

## Key Innovations

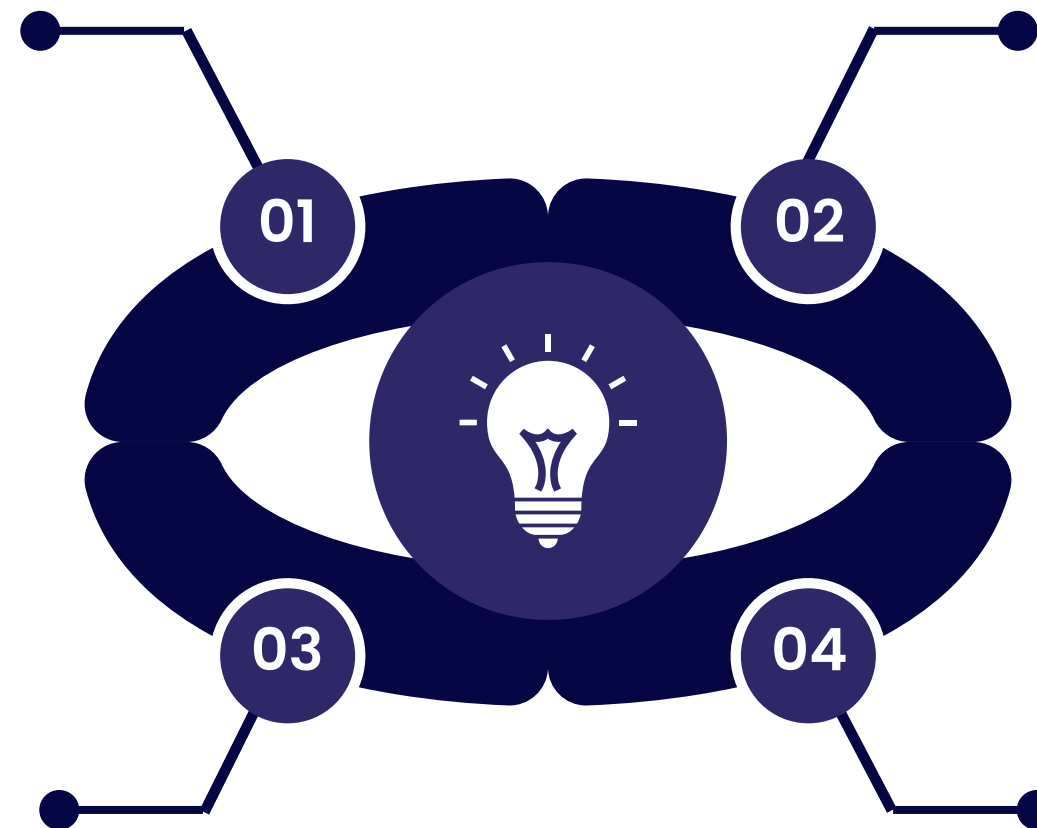
- Adaptive night-vision processing
- Real-time violation detection
- Automated license plate recognition
- Integrated tracking system



# Key Features of Project

**Adaptive Traffic Light State Detection at Night:** Utilizes advanced computer vision techniques to dynamically detect traffic light states in low-light conditions, ensuring accurate and reliable performance during nighttime.

**License Plate Extraction from Night-Time Images:** Leverages state-of-the-art image processing methods to extract license plate regions from nighttime images, addressing challenges like low visibility and glare from headlights.



## **Stop Line Detection System:**

Employs robust algorithms to identify stop lines on roads, even under challenging lighting scenarios. This ensures that vehicles are penalized only when they cross clearly marked boundaries.

## **License Plate Processing and Violation Management**

The system uses **PyTesseract OCR** for accurate recognition of license plate text from captured images. Penalized vehicle plates are displayed in real-time, providing immediate feedback and transparency for both authorities and drivers. Additionally, a **planned MySQL database** integration will enable seamless storage and retrieval of violation records, streamlining long-term data management and enforcement processes.



# Dependencies & Requirements

## OpenCV

- Capturing and preprocessing video frames.
- Converting color spaces (e.g., BGR to HSV) for adaptive traffic light detection.
- Performing operations like edge detection (Canny) and contour detection for license plate extraction.
- Drawing masks, bounding boxes, and overlaying real-time text on video frames.



## Numpy and PyTesseract:

**PyTesseract** was used for Optical Character Recognition (OCR) for converting license plate images into text.

**Numpy** was used for Efficient handling of image matrices (pixel-level operations)., processing binary masks for color and contour detection. Also used for calculating statistical properties (e.g., area of detected regions) to identify license plates or stop lines..

## MySQL Connector & PIL (Python Imaging Library)

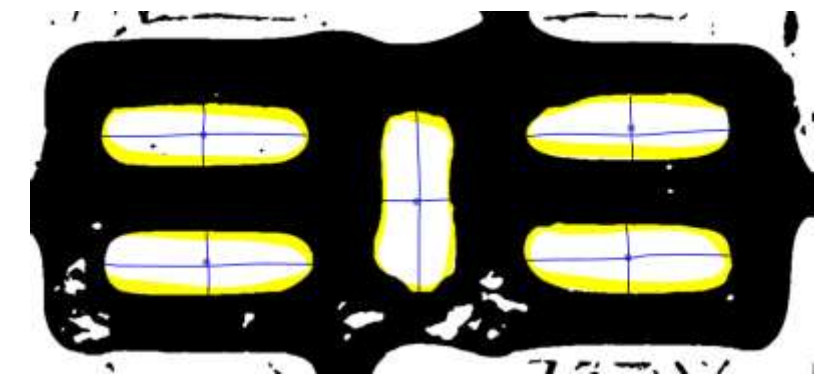
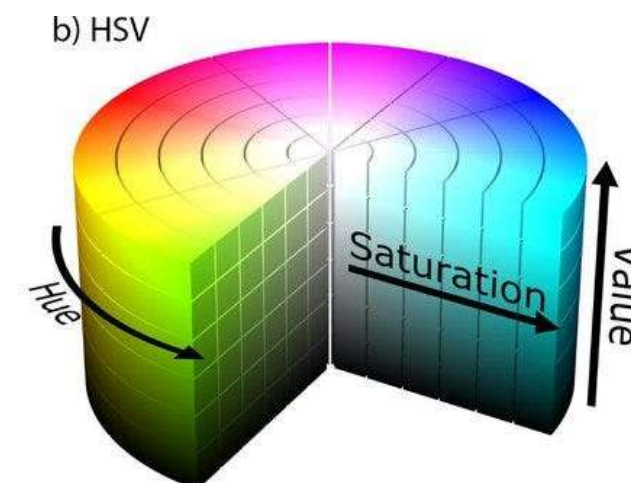
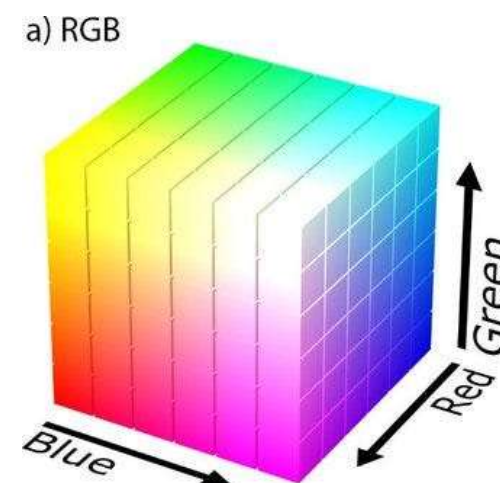
- **MYSQL Connector** was used for storing records of penalized vehicles, including license plate numbers, timestamps, and violation details. Facilitating query-based retrieval for managing historical data or generating reports.
- **PIL** was used for enhancing image quality (e.g., brightness and contrast adjustments) before feeding them to OCR systems. Resizing, cropping, and converting image formats to suit different processing stages.

# Adaptive Traffic Light State Detection at Night

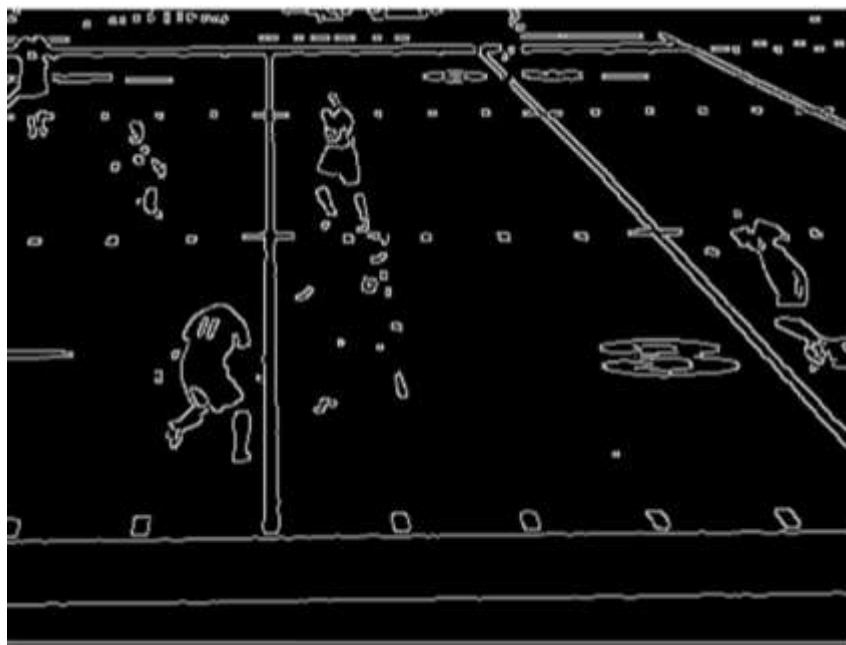
**Objective:** Detect the state of a traffic light (Red, Yellow, Green) in video frames captured at night.

## Techniques Used:

- **HSV (Hue, Saturation, Value) Color Space:** Frames are converted from BGR to HSV for better color segmentation. HSV separates color (Hue) from intensity (Value), making it ideal for detecting specific colors in low-light conditions.
- **Color Masks:** Threshold ranges are defined for red, yellow, and green lights. For example: Red:  $[0, 120, 70]$  to  $[10, 255, 255]$ , Yellow:  $[20, 100, 100]$  to  $[30, 255, 255]$ . Binary masks are created to isolate regions matching these colors.
- **Mask Analysis:** Using `cv2.countNonZero(mask)`, the system checks for significant color presence in the region of interest (ROI).
- **Real-Time Feedback:** Detected states are overlaid on the video frame using OpenCV's text overlay features.



Concept of `cv2.countNonZero(mask)`



Concept of Hough Transform



(d) Traffic congestion

# Stop Line Detection System

**Objective:** Detect the stop line in traffic scenes to correlate with traffic light states.

## Techniques Used:

- **Area Segmentation with Line Equations** :Equations define boundaries for the area around the stop line. *Example:  $y = slope * x + intercept$ .* Areas outside these boundaries are masked to simplify detection
- **Edge Detection and Hough Transform:** Used Canny Edge Detection for highlighting edges, reducing the image to essential line structures.
- **Hough Transform:** Detects lines in the binary edge map. Parameters like minLineLength and maxLineGap ensure accurate and connected line detection.
- **Line Averaging:** Detected lines across frames are averaged using deques, ensuring stable and consistent detection.



# License Plate Extraction from Night-Time Images

**Objective:** Isolate the license plate region in images for further processing.

## Techniques Used:

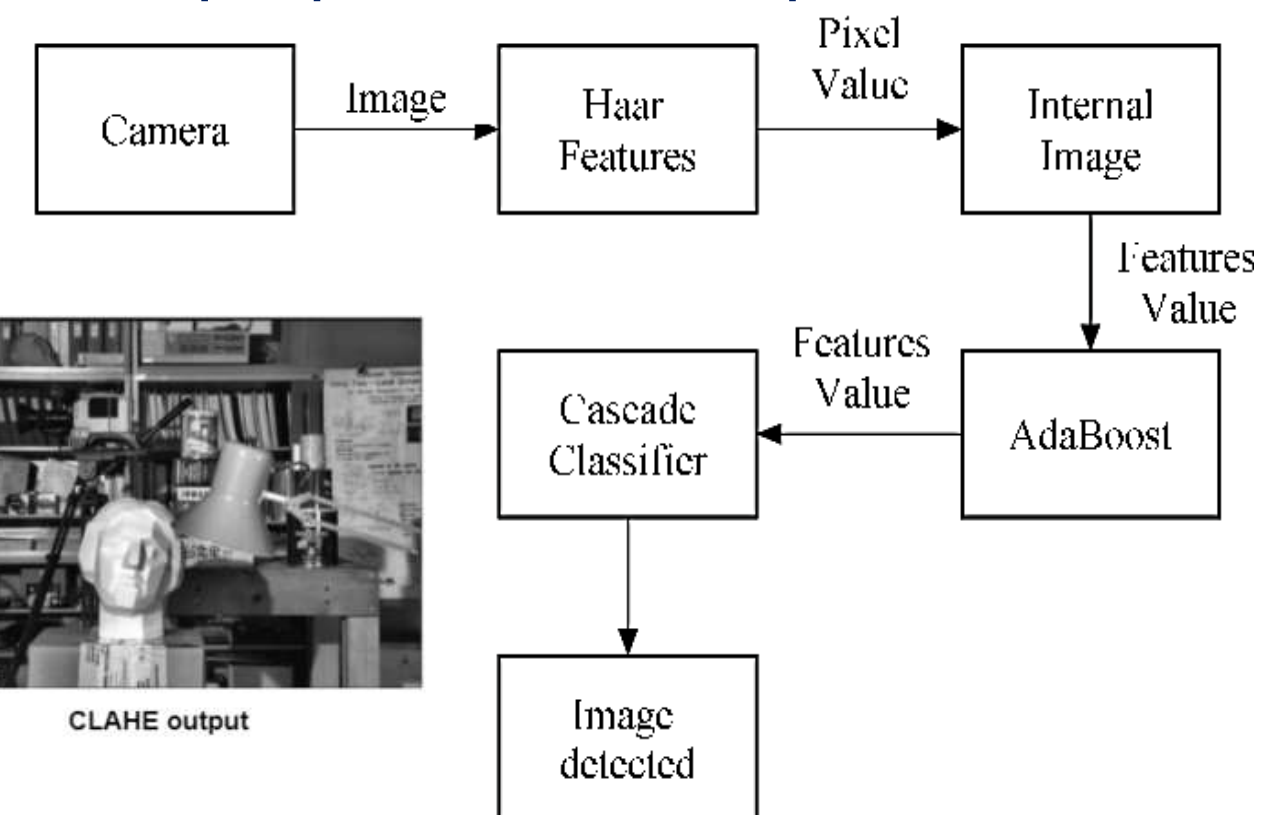
- **Region of Interest (ROI):** Bounding boxes are applied to focus on potential license plate areas.
- **Image Preprocessing:** Used *CLAHE* (*Contrast Limited Adaptive Histogram Equalization*) for enhancing contrast in low-light images, making text and edges more distinguishable.
- **Edge Detection (Canny):** Detects edges in the image, helping to locate rectangular structures like license plates.
- **Contour Detection:** Filters contours based on size and aspect ratio to pinpoint license plates.
- **Haar Cascade Classifier:** The cropped *grayscale image* is then processed using a Haar cascade classifier, a traditional non-deep learning approach proficient in pattern recognition. This classifier has been pre-trained to identify license plate patterns, returning the locations of possible license plates as rectangles.



Original Image



CLAHE output

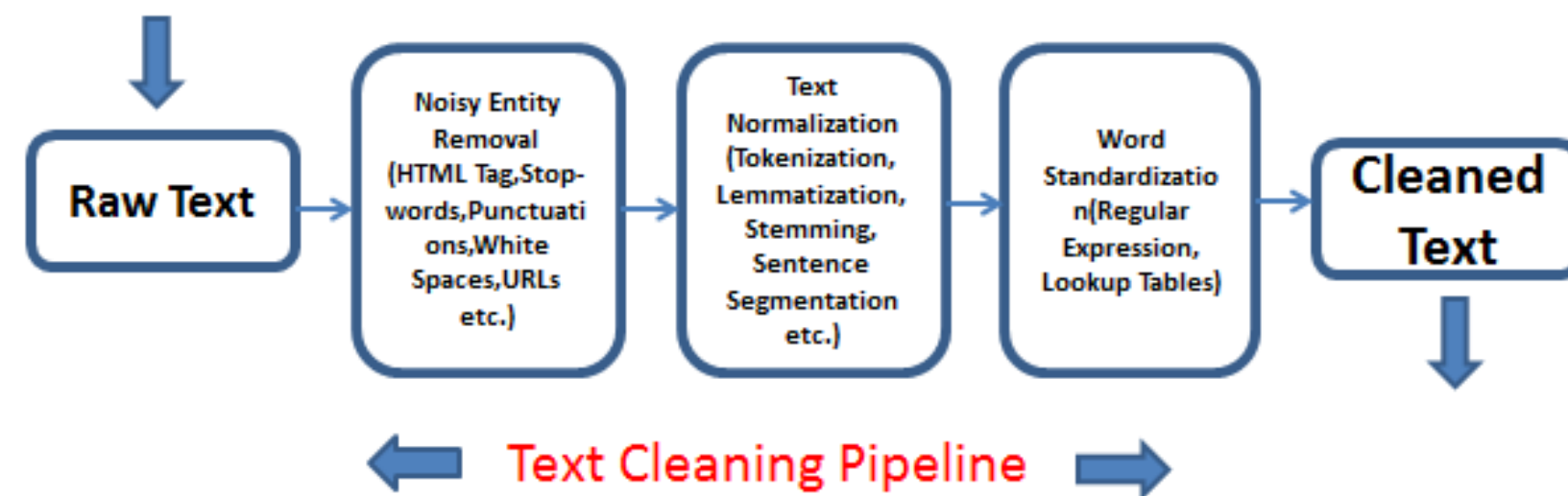


# License Plate Processing and Violation Management

**Objective:** Extract text from license plates and manage penalized vehicle records.

## Techniques Used:

- **OCR with PyTesseract:** PyTesseract, an OCR engine, converts the license plate image into text. Preprocessing steps like thresholding, resizing, and denoising improve recognition accuracy.
- **Text Regularization:** Regex filters out unwanted characters, ensuring only alphanumeric data is retained.
- **Real-Time Display:** Penalty information is dynamically displayed on the processed video frame.
- **Database Integration:** A planned feature involves storing violations in a MySQL database for efficient record-keeping.





# Key Findings

## ❑ Accuracy of Adaptive Traffic Light Detection:

- The HSV-based color detection method works effectively at night, even under low visibility, by isolating traffic light colors.
- Challenges include handling glare from other light sources, requiring further fine-tuning of thresholds.

## ❑ Stop Line Detection Efficiency:

- Edge detection and Hough Transform algorithms accurately identify stop lines, providing reliable inputs for monitoring violations.
- Detection becomes less accurate in frames with heavy shadows or overlapping vehicles.

## ❑ License Plate Extraction and OCR Performance:

- Preprocessing techniques like CLAHE and thresholding significantly improve OCR results for text extraction from license plates.
- Challenges include recognizing non-standard fonts and plates with severe reflections or damage.

## ❑ Real-Time Processing Capabilities:

- The system effectively processes video frames and displays penalized plates in real-time.
- Optimizations, such as region-based processing, reduce computational load but may miss plates on very fast-moving vehicles.

## ❑ Planned Database Integration:

- MySQL integration will enhance the system's ability to store and manage violations, but its absence in the current stage limits historical analysis.

The project demonstrates a strong proof of concept for automated traffic monitoring at night. Future improvements should focus on handling edge cases, like occlusions, dynamic weather conditions, and complex traffic scenarios.

# Thank You!

