



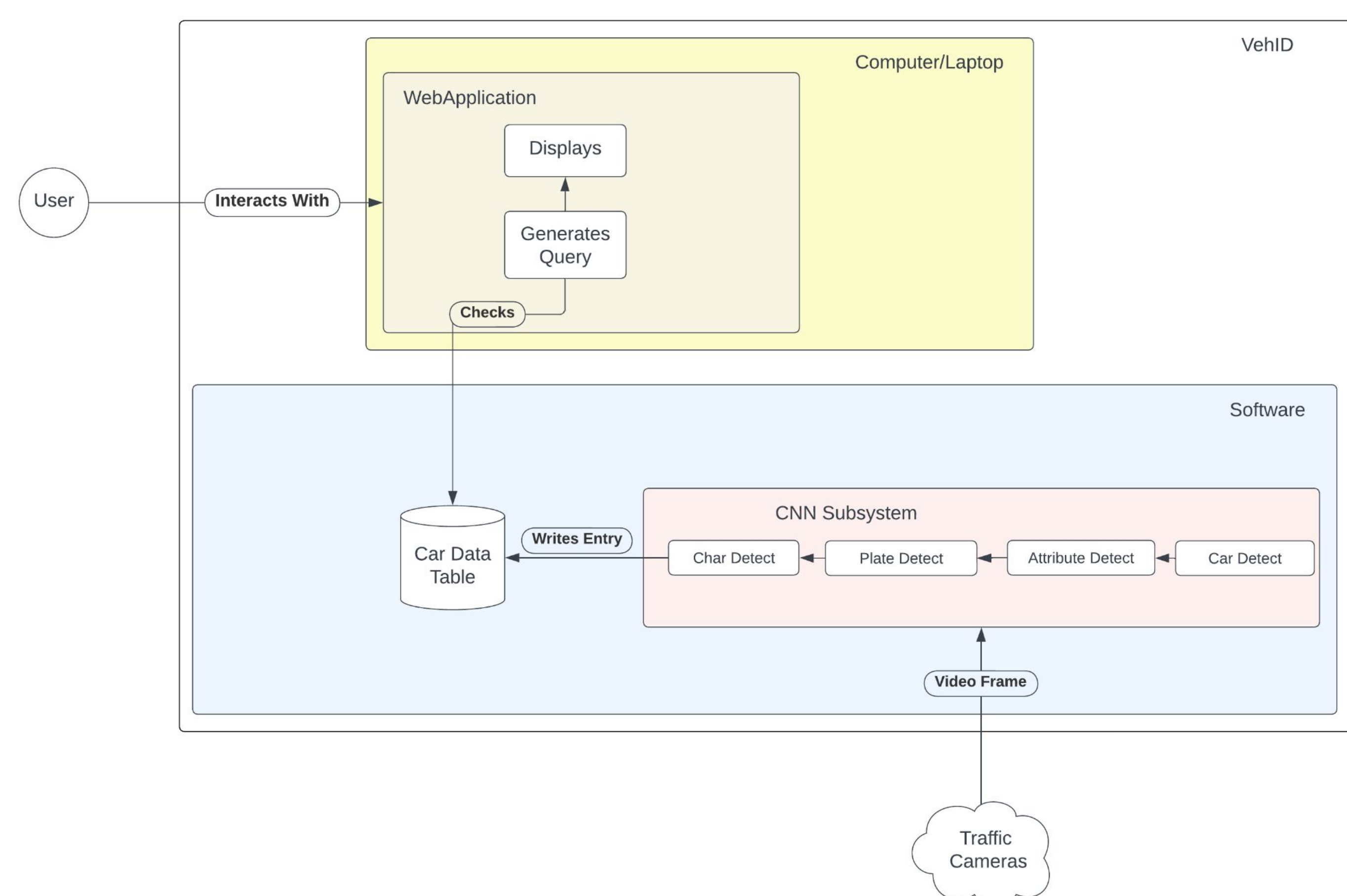
VehID

Remington Greko, Spencer Hirsch, Thomas Johnson, Alexis Nagle
 Faculty Advisor(s): Marius Silaghi, Dept. of Electrical Engineering and Computer Science,
 Florida Institute of Technology

Goal

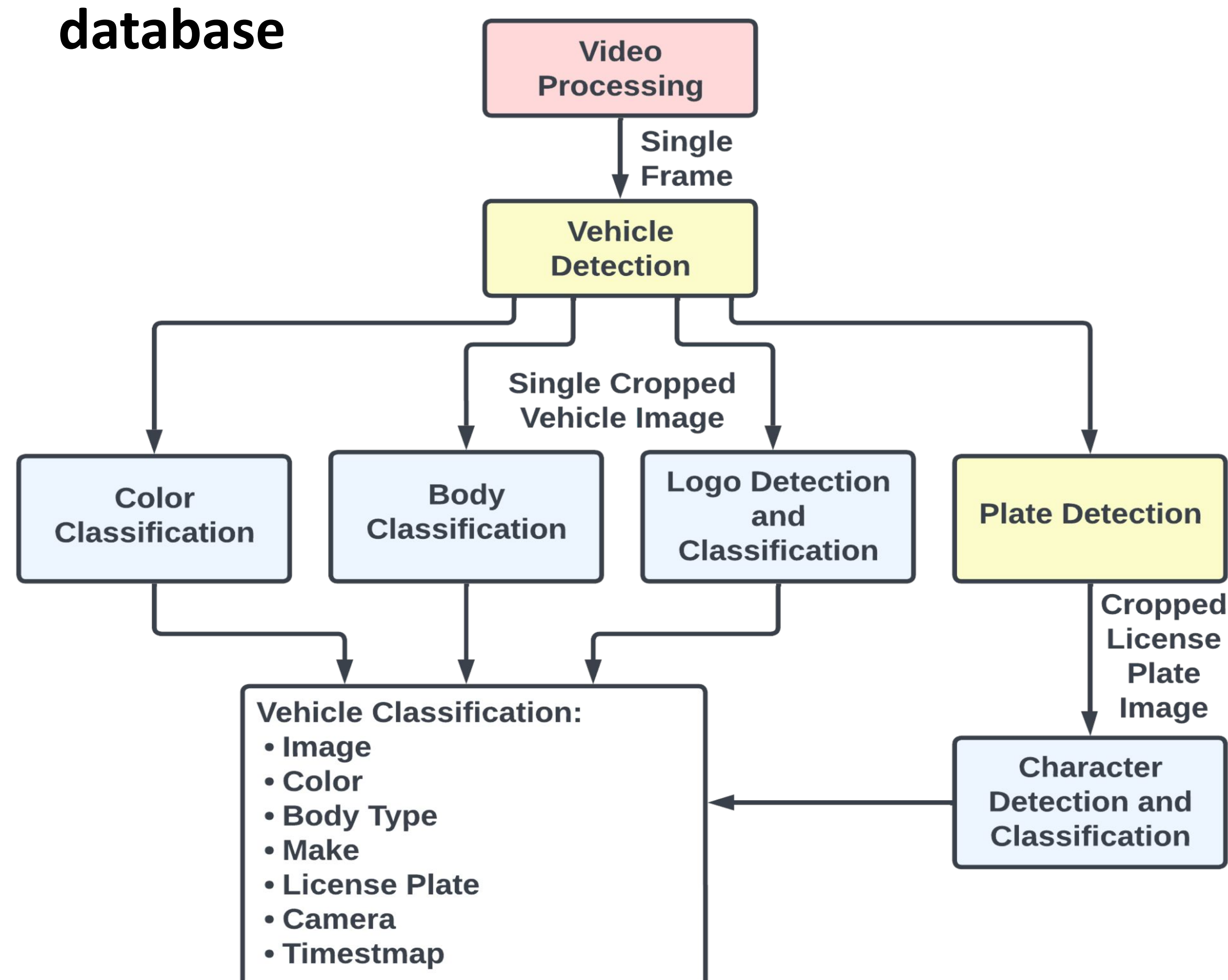
To improve public safety by utilizing machine learning to recognize vehicles based upon a variety of characteristics such as color, body type, make, and/or license plate.

Design



Neural Network (NN) Subsystem

- Utilized pre-existing NN architectures
 - MiniVGG (Body and Color Classifications)
 - VoloV8 (Vehicle, Logo, Plate, and Character Detections)
- Exports results to a JSON file for use in the database



Motivation

- Aid in AMBER alerts, stolen vehicles, and criminal offenses, which tends to rely on pure human interaction to spot and report the specified vehicles.
- Existing automation only identifies license plate numbers and is not beneficial when only other characteristics are available.

Database

- Car database: Stores entries of vehicles detected by the NN subsystem.
 - Includes image path, color, body type, make, plate, camera ID, and timestamp.

Web Application

- Allows for users to interact with the data contained within the databases.
- User functionalities:
 - Perform queries to search for vehicles with desired characteristics
 - Filter detected vehicle entries to be displayed on its respective web page
 - Edit and delete detected vehicle car entries
 - View saved images for each car entry for manual verification
- Example Query for any detected Grey-Silver vehicles and corresponding results displayed in the modal.

Select Vehicle Color:

Select Vehicle Body:

Select Vehicle Make:

License Plate Number:

Location:

ID	Color	Body	Make	License	Location	Image
5	grey-silver	Hatchback	Chrysler		Camera1	
6	grey-silver	Convertible	Chrysler	KTED7Z	Camera1	
8	grey-silver	Convertible	KTE47E	Camera1		
9	grey-silver	SUV	subaru	BSDJRV	Camera1, 05:00:25	
10	grey-silver	SUV	ford	BZR	Camera1, 05:00:26	
12	grey-silver	Pick-Up	GSZ	Camera1, 05:00:55		
14	grey-silver	Pick-Up	Z4W	Camera1, 05:02:10		
17	grey-silver	VAN	kia	U2ABRV	Camera2, 05:00:12	
18	grey-silver	SUV	subaru	ZRY	Camera2, 05:00:13	
21	grey-silver	Hatchback		GSBZ6	Camera2, 05:00:42	
22	grey-silver	Pick-Up		Z495M	Camera2, 05:02:00	

Implementation

- Neural Network Subsystem Tools:
 - Utilized various Python libraries
 - OpenCV
 - Tensorflow: Keras, MiniVGG model
 - Ultralytics: YoloV8 models
- Web Application Tools:
 - HTML/CSS
 - JavaScript
 - Node

Evaluation Results

- Neural Network Subsystem Evaluation:
 - 95% positive vehicle predictions
 - 45% positive color predictions
 - 32% positive body type predictions
 - 47% positive make predictions (out of the ones that had predictions)
 - 71% had 1 or more attributes correctly predicted
 - 28% had 2 or more attributes correctly predicted
- Web Application Evaluation:
 - Accurate and timely query results
 - Accurate and timely filter results
 - Timely table population
 - Sleek and intuitive web application design

Limitations

- Image/Video Quality
- Weather/Environmental Factors

Improvements

- Further training and tuning on select models to improve individual performances.
- Implement web scraping to automatically populate the database with queries from sources such as AMBER alerts.
- Host web application so queries can be ran repeatedly at set time increments
- Incorporate into a network of existing cameras.