

AUTOMATED TRAFFIC COUNTING DATA COLLECTION AND ANALYSIS

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INTRODUCTION

- The increase in vehicles purchased by households over the years cause a high volume of vehicles on the road has caused traffic congestion.
- Traffic data collection analysis to extract traffic information which would be useful for evaluating the quality of transportation and improve traffic control.
- This project proposed an automated traffic counting data collection and analysis algorithm that is able to use computer vi
 - to count vehicles
 - to measure the speed of vehicles
 - to classify vehicles into different categories using the power of deep learning and Al.

METHODOLOGY

Raspberry Pi 4 and RaspiCam V2 captures video of vehicles on the road

Frames (image) is extracted from the video one by one to the algorithm

Remove background using OpenCV's MOG2 Background Subtraction Apply thresholding, morphology opening and closing to produce a clean image

OpenCV's findContour is used to detect and track vehicle

If new vehicle register, else update coordinate of old vehicle

Direction of the vehicle can be determined

Distance travelled (pixels) of vehicle is calculated
 Distance travelled in pixels is converted into distance travelled in meters
 Speed of vehicles = Distance/Time

Pre-trained object detection model is trained using images from videos collected (Transfer Learning) Frames is inserted into the object detection model

•Output is image annotated with detection box around detected object and class of object

Time and date of vehicle recorded
Direction, vehicle class and speed of vehicle is recorded
Csv format

OBJECTIVES

To capture good quality video recording of vehicles on the road.



To detect, track and count vehicles in the video.



To calculate the **speed** of every vehicle passing in the video.



To **classify** the vehicles in the video into cars, motorcycles using Al.



RESULTS AND DISCUSSION

Vehicle Counting



Figure 1: Bar chart of actual number of vehicles vs number of vehi-

3.18181818 Morning 235 261 Afternoon 161 Night 126 Night 162

Figure 2: table of actual number of vehicles vs number of vehicles counted by algorithm with accuracy

CONCLUSION

The project fulfills the objectives stated. Vehicle counting, speed measurement and vehicle classification achieves high accuracy and above when scene is well lit. When the scene is totally dark, vehicle counting still performs decently, while classification performs poorly. The algorithm is able to collect different types of data useful for traffic analysis.

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Vehicle Classification

Day	Time	Actual Cars	Predicted Cars	Accuracy (%)
1	Morning	171	156	91.22807018
1	Afternoon	121	103	96.42857
1	Night	124	77	62.09677419
2	Morning	172	163	94.76744186
2	Afternoon	85	81	95.29411765
2	Night	91	28	30.76923077

Figure 3: Table of actual number of cars vs number of cars classified by the object detection model

Data Logging

/ear	Month	Day	Time	ObjectID	Direction	Vehicle Class	Speed(km/h)
2022	4	3	17:24:32	0	Right	car	26.33
2022	4	3	17:24:38	1	Right	car	29.09
2022	4	3	17:24:43	2	Right	car	33.76
2022	4	3	17:24:53	3	Right	car	22.25
2022	4	3	17:24:56	4	Right	car	19.24
2022	4	3	17:24:59	5	Right	car	17.32
2022	4	3	17:25:02	6	Right	car	14.19
2022	4	3	17:25:06	7	Right	car	22.93
2022	4	3	17:25:11	8	Right	car	15.18
2022	4	3	17:25:16	10	Right	car	28.17
2022	4	3	17:25:20	11	Right	car	19.52
2022	4	3	17:25:55	12	Right	car	37.87
2022	4	3	17:26:01	13	Left	car	52.65
2022	4	3	17:26:03	14	Right	motorcycle	29.13
2022	4	3	17:26:07	15	Right	car	36.24
2022	4	3	17:26:09	17	Left	motorcycle	33.64
2022	4	3	17:26:20	18	Right	car	31.85
2022	4	3	17:26:33	19	Left	car	21.21
2022	4	3	17:26:44	21	Left	car	64.98

Figure 5: Data collected by the software

		Actual	Predicted	
Day	Time	Motorcycles	Motorcycles	Accuracy (%)
1	Morning	49	40	81.63265306
1	Afternoon	40	38	95
1	Night	58	10	17.24137931
2	Morning	57	41	71.92982456
2	Afternoon	33	33	100
2	Night	34	5	14.70588235

Figure 4: Table of actual number of motorcycles vs number of motorcycles classified by the object detection model

The algorithm is tested using 6 videos recorded at different times. The video in the morning and afternoon are brightly lit. In the first night there is still some light in the sky, while in the second night the scene is totally dark with one road lamp illuminating the road. Vehicle counting, speed measurement and vehicle classification achieves high accuracy of 90% and above when scene is well lit. When the scene is totally dark, vehicle counting accuracy drops to around 60% to 76%, while classification accuracy ranges from 30% to 60% for cars and 14% to 17% for motorcycles. The algorithm is able to record the date, time, direction, type of vehicle and speed of vehicles that pass by.