

**AIoT POTHOLE DETECTION SYSTEM  
USING YOLOV5**



**UNIVERSITI TEKNIKAL MALAYSIA MELAKA**



AIoT POTHOLE DETECTION SYSTEM  
USING YOLOV5

THIVEYA A/P MAHENDRAN



اونيورسيتي تیکنیکل ملیسيا ملاک

This report is submitted in partial fulfillment of the requirements for the Bachelor of Computer Science (Artificial Intelligence) with Honours.

FACULTY OF INFORMATION AND COMMUNICATION TECHNOLOGY  
UNIVERSITI TEKNIKAL MALAYSIA MELAKA

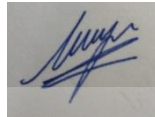
2023

## DECLARATION

I hereby declare that this project report entitled

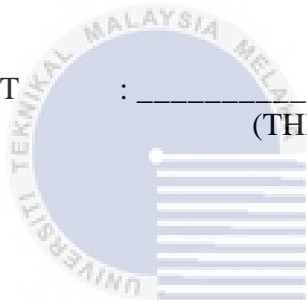
### **AIoT POTHOLE DETECTION SYSTEM USING YOLOV5**

is written by me and is my own effort and that no part has been plagiarized  
without citations.



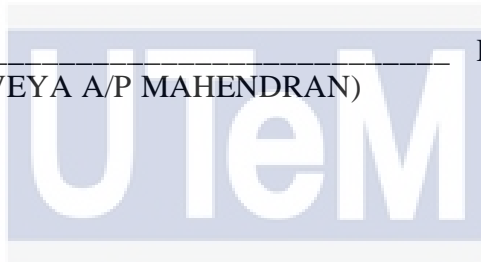
STUDENT

:



(THIVEYA A/P MAHENDRAN)

Date : 29/9/2023



اونيورسيتي تيكنيكل مليسيا ملاك

UNIVERSITI TEKNIKAL MALAYSIA MELAKA

I hereby declare that I have read this project report and found  
this project report is sufficient in term of the scope and quality for the award of  
Bachelor of Computer Science (Artificial Intelligence) with Honours.

SUPERVISOR

:



(DR. NUR ZAREEN ZULKARNAIN)

Date : 27/9/2023

## DEDICATION

My FYP supervisor, DR. Nur Zareen Zulkarnain deserves special thanks for her unwavering support throughout the project's completion, her patience in guiding me through every phase of the project and for inspiring and encouraging me while I worked on this project. I am grateful for the assistance my supervisor provided. I would like to dedicate to my cherished parents and siblings, who have continuously given me the willpower, support, and motivation to complete my project during my university years. Finally, I want to thank my friends for their unwavering support and assistance over the past years of my university life. Lastly, I would like to thank The Almighty God who has given me the strength and willpower to complete this project.



## ACKNOWLEDGEMENTS

I would like to thank Dr Nur Zareen Zulkarnain for giving me assistance to complete this project successfully. Without her direction and assistance, I wouldn't be able to do this project. I also want to express my gratitude to my parents, who have supported and inspired me throughout my quest. In addition, I would like to thank my cherished family for their unwavering concern and spiritual guidance throughout my entire life, especially during my pursuit of this bachelor's degree. I'd like to use this chance to express my gratitude to my best friends for their constant support and assistance during this project. They have been there for me ever since I started my undergraduate studies and guided me during my hard times in both studies and in real life. I would also like to thank God for blessing me with an education and giving me strength to go through my tough times.



## ABSTRACT

The AIoT Pothole Detection System using YOLOv5 is an innovation that is made to help the society. The elements implemented in the project are artificial intelligence primarily focusing on object detection, IoT device integration and YOLOv5 algorithm. The objectives of this innovation is to implement object detection to detect potholes present on the road, measure the performance of two different YOLO models detection accuracy for pothole detection and evaluate the performance of the chosen YOLO model. The problems that can be solved from this innovation are plenty and they are to improve road safety of drivers and riders providing them enhanced driving experience, reducing damage to vehicles and provides efficient road maintenance helping reduce infrastructure damage. In addition to that, this project is a way to embody the state-of-the-art potentials like IoT (Internet of Things) and AI (Artificial Intelligence) technologies, especially incorporating them and producing futuristic innovations. Pothole detection system using YOLOv5 deployed to a Raspberry Pi 4 offers a holistic solution to a common urban infrastructure challenge and has a lot to contribute to the overall well-being and safety of communities.

## ABSTRAK

“AIoT Pothole Detection System” menggunakan YOLOv5 adalah satu inovasi yang dibuat untuk membantu masyarakat. Elemen yang dilaksanakan dalam projek ini ialah kecerdasan buatan yang memfokuskan terutamanya pada pengesanan objek, penyepaduan peranti IoT dan algoritma YOLOv5. Objektif inovasi ini adalah untuk melaksanakan pengesanan objek untuk mengesan jalan berlubang, mengukur prestasi dua model YOLO yang berbeza ketepatan pengesanan untuk pengesanan lubang dan menilai prestasi model YOLO yang dipilih. Masalah yang boleh diselesaikan daripada inovasi ini adalah banyak dan ia adalah untuk meningkatkan keselamatan jalan raya pemandu dan penunggang memberikan mereka pengalaman memandu yang lebih baik, mengurangkan kerosakan kenderaan dan menyediakan penyelenggaraan jalan yang cekap membantu mengurangkan kerosakan infrastruktur. Di samping itu, projek ini adalah satu cara untuk menjelmakan potensi terkini seperti teknologi IoT (Internet of Things) dan AI (Kecerdasan Buatan), terutamanya menggabungkannya dan menghasilkan inovasi futuristik. Sistem pengesanan berlubang menggunakan YOLOv5 yang digunakan pada Raspberry Pi 4 menawarkan penyelesaian holistik kepada cabaran infrastruktur bandar yang sama dan mempunyai banyak perkara untuk menyumbang kepada kesejahteraan dan keselamatan keseluruhan komuniti.



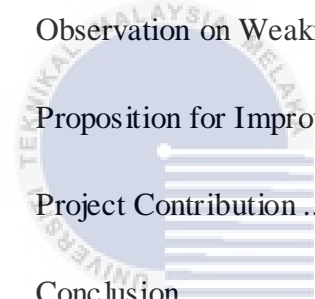
## TABLE OF CONTENTS

<b>DECLARATION.....</b>	<b>II</b>
<b>DEDICATION.....</b>	<b>III</b>
<b>ACKNOWLEDGEMENTS.....</b>	<b>IV</b>
<b>ABSTRACT.....</b>	<b>V</b>
<b>ABSTRAK.....</b>	<b>VI</b>
<b>TABLE OF CONTENTS.....</b>	<b>VII</b>
<b>LIST OF TABLES.....</b>	<b>XI</b>
<b>LIST OF FIGURES.....</b>	<b>XII</b>
<b>LIST OF ABBREVIATIONS.....</b>	<b>XV</b>
<b>CHAPTER 1: INTRODUCTION.....</b>	<b>16</b>
1.1 Introduction .....	16
1.2 Problem Statement(s) .....	17
1.3 Objective(s) .....	17
1.4 Scope.....	18
1.5 Project Significance.....	18
1.6 Expected Output.....	19
1.7 Conclusion .....	19
<b>CHAPTER 2: LITERATURE REVIEW AND PROJECT METHODOLOGY</b>	<b>20</b>
2.1 Introduction .....	20
2.2 Facts and Findings.....	20
2.2.1 Domain.....	22
2.2.2 Existing System .....	23

2.2.3	Techniques chosen for Pothole Detection System (PDS) .....	32
2.3	Project Methodology .....	39
2.3.1	Data Collection .....	39
2.3.2	Data Pre-processing .....	39
2.3.3	Model Selection .....	40
2.3.4	Model Training .....	40
2.3.5	Model Evaluation .....	41
2.4	Project Requirements .....	41
2.4.1	Software Requirement(s) .....	42
2.4.2	Hardware Requirement(s) .....	42
2.5	Project Schedule and Milestones .....	43
2.6	Conclusion .....	45
<b>CHAPTER 3: ANALYSIS .....</b>		<b>46</b>
3.1	Introduction .....	46
3.2	Problem Analysis .....	46
3.3	Requirement analysis .....	48
3.3.1	Data Requirement .....	48
3.3.2	Functional Requirement .....	54
3.3.3	Non-functional Requirement .....	54
3.3.4	Others Requirement .....	55
3.4	Conclusion .....	57
<b>CHAPTER 4: DESIGN .....</b>		<b>58</b>
4.1	Introduction .....	58
4.2	High-Level Design .....	58

4.2.1	System Architecture for Pothole Detection System.....	59
4.2.2	User Interface Design for PDS.....	62
4.3	Detailed Design.....	64
4.3.1	Data Augmentation .....	64
4.3.2	Custom Trained YOLOv5 Model.....	66
4.3.3	Custom Trained YOLOv8s Model.....	69
4.3.4	Detection .....	72
4.4	Conclusion.....	72
<b>CHAPTER 5: IMPLEMENTATION.....</b>		<b>73</b>
5.1	Introduction .....	73
5.2	Software Development Environment Setup .....	73
5.3	Configuration Management .....	75
5.3.1	Software Configuration Management.....	75
5.3.2	Hardware Configuration Management .....	76
5.4	Implementation Status .....	76
5.5	Conclusion.....	77
<b>CHAPTER 6: TESTING.....</b>		<b>78</b>
6.1	Introduction .....	78
6.2	Test Plan.....	78
6.2.1	Test Organization.....	78
6.2.2	Test Environment.....	79
6.2.3	Test Schedule .....	80
6.3	Test Strategy.....	81
6.3.1	Classes of Test.....	81

6.4	Test Implementation.....	81
6.4.1	Experimental/ Test Description.....	82
6.4.2	Test Data.....	87
6.5	Test Results and Analysis .....	87
6.5.1	Results of Real World Accuracy Testing from Laptop Webcam...	87
6.5.2	Results of Installation and Setup Verification of Raspberry Pi 4...	90
6.5.3	Results of Real World Accuracy Testing from Raspberry Pi 4 connected with Webcam.....	93
6.6	Conclusion.....	99
<b>CHAPTER 7: CONCLUSION .....</b>		<b>100</b>
7.1	Observation on Weaknesses and Strengths .....	100
7.2	Proposition for Improvement .....	101
7.3	Project Contribution .....	101
7.4	Conclusion.....	101
<b>REFERENCES.....</b>		<b>102</b>



اونيورسيتي تیکنیکل ملیسيا ملاک

UNIVERSITI TEKNIKAL MALAYSIA MELAKA

## LIST OF TABLES

<b>Table 1:</b>	<b>Table of performance results of the algorithms using previously determined optimal parameters</b>	<b>26</b>
<b>Table 2:</b>	<b>Performance measure for all the techniques used.</b>	<b>28</b>
<b>Table 3:</b>	<b>Software Requirements and specifications of PDS</b>	<b>44</b>
<b>Table 4:</b>	<b>Hardware Requirements and specifications of PDS</b>	<b>44</b>
<b>Table 5:</b>	<b>Functional Requirements in PDS</b>	<b>54</b>
<b>Table 6:</b>	<b>Non- Functional Requirements in PDS</b>	<b>55</b>
<b>Table 7:</b>	<b>Python libraries and their functionalities.</b>	<b>56</b>
<b>Table 8:</b>	<b>Software requirements and their functionalities.</b>	<b>57</b>
<b>Table 9:</b>	<b>Hardware requirements and their functionalities.</b>	<b>57</b>
<b>Table 10:</b>	<b>Software development environment setup</b>	<b>74</b>
<b>Table 11:</b>	<b>Implementation status of PDS</b>	<b>77</b>
<b>Table 12:</b>	<b>Roles and Responsibilities</b>	<b>78</b>
<b>Table 13:</b>	<b>Test Environment</b>	<b>80</b>
<b>Table 14:</b>	<b>Test Schedule</b>	<b>80</b>
<b>Table 15:</b>	<b>Classes of Test</b>	<b>81</b>
<b>Table 16:</b>	<b>Real World Accuracy Testing from Laptop Webcam</b>	<b>83</b>
<b>Table 17:</b>	<b>Installation and Setup Verification of Raspberry Pi 4</b>	<b>84</b>
<b>Table 18:</b>	<b>Real World Accuracy Testing from Raspberry Pi 4 connected with Webcam</b>	<b>86</b>

## LIST OF FIGURES

<b>Figure 1:</b>	<b>Statistics of accidents involving potholes in Malaysia</b>	<b>21</b>
<b>Figure 2:</b>	<b>Raw data and 3D model data of pothole</b>	<b>24</b>
<b>Figure 3:</b>	<b>Example of edge detection and K-mean clustering image segmentation</b>	<b>28</b>
<b>Figure 4:</b>	<b>Graph visuals on the performance (loss and mAP_0.5) of Yolov4 (a), Yolov4-tiny (b) and Yolov5 (c)</b>	<b>31</b>
<b>Figure 5:</b>	<b>Example of bounding box representation</b>	<b>33</b>
<b>Figure 6:</b>	<b>YOLO algorithm techniques applied to produce the final detection results.</b>	<b>34</b>
<b>Figure 7:</b>	<b>The architecture of CNN model</b>	<b>35</b>
<b>Figure 8:</b>	<b>Flow of project methodology.</b>	<b>38</b>
<b>Figure 9:</b>	<b>Milestone of project</b>	<b>44</b>
<b>Figure 10:</b>	<b>Flowchart of pothole detection system</b>	<b>47</b>
<b>Figure 11:</b>	<b>Pothole Dataset</b>	<b>48</b>
<b>Figure 12:</b>	<b>Data annotations performed in Roboflow</b>	<b>49</b>
<b>Figure 13:</b>	<b>Diverse set of potholes</b>	<b>49</b>
<b>Figure 14:</b>	<b>Flipped pothole data augmentation</b>	<b>50</b>
<b>Figure 15:</b>	<b>Dataset split into training, validation and testing sets.</b>	<b>50</b>
<b>Figure 16:</b>	<b>Dataset divided in Roboflow for train, validate and test</b>	<b>51</b>
<b>Figure 17:</b>	<b>Example code snippet when YOLO model is exported from Roboflow</b>	<b>51</b>
<b>Figure 18:</b>	<b>Python libraries in Google Colab.</b>	<b>52</b>
<b>Figure 19:</b>	<b>Model trained with 300 epochs in Google Colab.</b>	<b>52</b>
<b>Figure 20:</b>	<b>Inference run displayed on test images in Google Colab.</b>	<b>53</b>
<b>Figure 21:</b>	<b>Code to zip YOLO model folder</b>	<b>53</b>
<b>Figure 22:</b>	<b>YOLO folders containing important information.</b>	<b>53</b>
<b>Figure 23:</b>	<b>A detailed flow structure of the pothole detection system.</b>	<b>59</b>
<b>Figure 24:</b>	<b>Bounding box tool used for annotations and label classes as 'potholes'</b>	<b>60</b>
<b>Figure 25:</b>	<b>Torch hub load code for YOLOv5.</b>	<b>61</b>
<b>Figure 26:</b>	<b>openCV code snippet</b>	<b>61</b>
<b>Figure 27:</b>	<b>Code snippet of YOLOv8 testing in Jupyter Notebook.</b>	<b>61</b>

<b>Figure 28:</b>	<b>Output design of PDS</b>	<b>63</b>
<b>Figure 29:</b>	<b>Flip Augmentation in Roboflow</b>	<b>64</b>
<b>Figure 30:</b>	<b>Brightness Augmentation in Roboflow</b>	<b>64</b>
<b>Figure 31:</b>	<b>Crop Augmentation in Roboflow</b>	<b>65</b>
<b>Figure 32:</b>	<b>Noise Augmentation in Roboflow</b>	<b>65</b>
<b>Figure 33:</b>	<b>Code snippet of YOLOv5 exporting format.</b>	<b>66</b>
<b>Figure 34:</b>	<b>Training code of YOLOv5</b>	<b>66</b>
<b>Figure 35:</b>	<b>Pothole validation of YOLOv5</b>	<b>67</b>
<b>Figure 36:</b>	<b>Precision-Recall curve graph of YOLOv5</b>	<b>67</b>
<b>Figure 37:</b>	<b>Confusion matrix of YOLOv5.</b>	<b>68</b>
<b>Figure 38:</b>	<b>Code snippet of YOLOv8s exporting format.</b>	<b>69</b>
<b>Figure 39:</b>	<b>Training code of YOLOv8s</b>	<b>69</b>
<b>Figure 40:</b>	<b>Pothole validation of YOLOv8s</b>	<b>70</b>
<b>Figure 41:</b>	<b>Precision-Recall curve graph of YOLOv5</b>	<b>70</b>
<b>Figure 42:</b>	<b>Confusion matrix of YOLOv5.</b>	<b>71</b>
<b>Figure 43:</b>	<b>GitHub platform used as code hosting platform for version control.</b>	<b>75</b>
<b>Figure 44:</b>	<b>Hardware Configuration of PDS</b>	<b>76</b>
<b>Figure 45:</b>	<b>Results of day time vision detection for YOLOv5.</b>	<b>87</b>
<b>Figure 46:</b>	<b>Results of day time vision detection for YOLOv8.</b>	<b>88</b>
<b>Figure 47:</b>	<b>Testing pothole using laptop webcam during daytime.</b>	<b>88</b>
<b>Figure 48:</b>	<b>Testing pothole using laptop webcam during night time with low light intensity.</b>	<b>89</b>
<b>Figure 49</b>	<b>Results of night time vision detection for YOLOv5.</b>	<b>89</b>
<b>Figure 50:</b>	<b>Results of night time vision detection for YOLOv8.</b>	<b>90</b>
<b>Figure 51:</b>	<b>Raspberry Pi 4 Model B</b>	<b>90</b>
<b>Figure 52:</b>	<b>Switch button for HDMI Display</b>	<b>91</b>
<b>Figure 53:</b>	<b>Upon switching the ON button, HDMI Display powers on the screen and shows Raspberry Pi 4 desktop.</b>	<b>91</b>
<b>Figure 54:</b>	<b>OpenCV has been installed upon entering command in terminal on Raspberry Pi 4 desktop.</b>	<b>92</b>

<b>Figure 55:</b>	<b>Command to install python3 on Raspberry Pi 4.</b>	<b>92</b>
<b>Figure 56:</b>	<b>Command to verify python and opencv has been installed in Raspberry Pi 4.</b>	<b>93</b>
<b>Figure 57:</b>	<b>Results of pothole detection on Raspberry Pi from front dashboard position during day time.</b>	<b>93</b>
<b>Figure 58:</b>	<b>Results of pothole detection on Raspberry Pi from front dashboard position during night time.</b>	<b>94</b>
<b>Figure 59:</b>	<b>Results of pothole detection on Raspberry Pi from side mirror position at day time.</b>	<b>94</b>
<b>Figure 60:</b>	<b>Results of pothole detection on Raspberry Pi from side mirror position at night.</b>	<b>95</b>
<b>Figure 61:</b>	<b>Results of pothole detection on Raspberry Pi at low speed driving.</b>	<b>95</b>
<b>Figure 62:</b>	<b>Results of pothole detection on Raspberry Pi at average speed driving.</b>	<b>96</b>
<b>Figure 63:</b>	<b>Confusion matrix of YOLOv5 model used for Raspberry Pi 4.</b>	<b>97</b>
<b>Figure 64:</b>	<b>Formulas of precision and recall</b>	<b>97</b>
<b>Figure 65:</b>	<b>Precision-Recall curve of YOLOv5 model used in Raspberry Pi 4.</b>	<b>98</b>
<b>Figure 66:</b>	<b>Results of YOLOv5 AI model deployed in Raspberry Pi 4.</b>	<b>99</b>



## LIST OF ABBREVIATIONS

<b>FYP</b>	-	<b>Final Year Project</b>
<b>AI</b>	-	<b>Artificial Intelligence</b>
<b>IoT</b>	-	<b>Internet of Things</b>
<b>AIoT</b>	-	<b>Artificial Intelligence Internet of Things</b>
<b>CNN</b>	-	<b>Convolutional Neural Network</b>
<b>RPI 4</b>	-	<b>Raspberry Pi 4</b>



اونيورسيتي تيكنيكل مليسيا ملاك

UNIVERSITI TEKNIKAL MALAYSIA MELAKA

## CHAPTER 1: INTRODUCTION

### 1.1 Introduction

Potholes are areas of road surface that have cracked, worn away, and eventually formed a hole. They start out as tiny cracks. If they're not fixed right away, they can grow. Potholes may be anywhere from a few inches wide and deep to a few feet wide and several inches deep.

Vehicles depend on a flat, smooth surface to move across city streets and highways. They can move swiftly, pleasantly, and safely as a result. A ride over a pothole will be unsteady and maybe dangerous. They might potentially mess with a car's wheel alignment in addition to harming the tyres.

Potholes can cause a variety of damages, both to vehicles and to the infrastructure of roads. Some of the potential damages that can result from hitting a pothole is damage to vehicle tires where the potholes can cause punctures, flats, or blowouts in tires. The impact can also cause tire alignment issues or damage to the suspension system.

Besides that, the wheels and rims can result in bent or cracked rims that need to be repaired or replaced due to hitting the potholes. A major damage happens mainly to the suspension system including the shocks, struts, and springs which can lead to expensive repairs and reduced driving comfort.

The undercarriage of a vehicle, including the exhaust system, oil pan, or transmission can also be damaged due to hitting a pothole and lastly potholes can also cause damage to the roads themselves, which can lead to costly repairs for municipalities and governments. The damage can include cracking, crumbling, and erosion of the road surface, as well as damage to underground utilities' and infrastructure.

Due to all this safety threatening reasons, a pothole detection system has been proposed to overcome the issues faced by drivers or riders on the road. This system is believed to detect pothole while they are driving on the road and gives the driver a warning like sign to avoid the potholes on the road.

## 1.2 Problem Statement(s)

The roads in Malaysia as we all know has many dreaded potholes and it is one of the many annoyances that many drivers experience. No matter how shallow or deep they are, driving over them is extremely unpleasant. These tiny cracks in the pavement can disrupt commuter traffic and are a significant factor in axle and suspension failure in automobiles. They may develop as a result of poor road construction and design, inclement weather, and inadequate drainage.

Potholes in the road can cause a great deal of damage to cars and other vehicles. They can puncture tyres and ruin vehicles' suspension and steering alignment. They can also cause scratched paintwork and marked bumpers from the impact of the car hitting the broken road. Driving into a pothole can also cause loss of control over the car, which can lead to accidents involving more than one vehicle.

## 1.3 Objective(s)

The project's main objective is to overcome the menacing pothole issues faced by drivers or motorists on the road by developing the pothole detections system (PDS) using image recognition. The main objectives of the system are as such:

1. To detect the potholes present on the road using computer vision (object detection).
2. To measure the performance of two different YOLO models detection accuracy for pothole detection.
3. To evaluate the performance of the chosen YOLO model.

## 1.4 Scope

The targeted users for this project is car drivers and bike riders including an vehicle drivers since vehicles nowadays can have its own dash camera fixed in it. Part of the scope is the modules to be developed and they are as such:

1. AI object detection for detecting the potholes on the road.
2. A system that can display alerts to drivers or riders.
3. Creating a beep alert sound using sensors.

## 1.5 Project Significance

The significance of having a pothole detection system can be important in many ways and one of it is improved safety. Potholes can pose a significant safety risk for drivers, particularly if they are not visible or if they are located in areas where drivers are not expecting them. PDS can help identify and locate potholes in real-time, allowing drivers to take appropriate action to avoid them.

Besides that, PDS is believed to reduce vehicle damage as potholes can cause significant damage to vehicles, including punctured tires, bent rims, and damage to the suspension system. By detecting and avoiding potholes, drivers can reduce the likelihood of incurring these types of costly repairs.

In addition to the project's significance, PDS can be cost saving by identifying and repairing potholes early where these costs can be minimized since potholes can cause significant damage to roads, resulting in costly repairs for municipalities and governments. With that being said, we can also have improved road maintenance where PDS can help municipalities and governments identify areas that require maintenance and repair more quickly and accurately. This can help prioritize maintenance efforts and ensure that resources are used more efficiently.

Overall, this pothole detection system (PDS) can help improve road safety, reduce vehicle damage, save costs, and improve road maintenance efforts. As such, it is an important tool for both drivers and municipalities alike.

## 1.6 Expected Output

The expected output of this project is as follows:

- The AI model correctly detects and recognizes the potholes present on the road.
- The system then alerts the driver or rider beforehand they can pass by the pothole from far (example 100m away).
- A beep sound to let the driver or rider avoid the pothole on the road.
- Implement PDS in real-time vehicles by default manufacturing and improvise vehicles safety on road.

UNIVERSITI TEKNIKAL MALAYSIA MELAKA

## 1.7 Conclusion

In conclusion, we have discussed what and how potholes form and how it leaves a menacing impact on people who use the road such as drivers, riders and pedestrians even. Besides that, we analyzed all the safety threatening problems of the potholes and how pothole detection system (PDS) can overcome all such problems in a simple way. Some of the expected outputs of PDS and how it functions were also discussed in this chapter.

## CHAPTER 2: LITERATURE REVIEW AND PROJECT METHODOLOGY

### 2.1 Introduction

In chapter 2, relevant project journals and research papers will be stated and discussed as they are used as citation in this project. This is a crucial step and it is done so that the objective of this project can be accomplished by first comprehending the necessary fundamental ideas, methods and techniques to be used.

The purpose of all these literature reviews are to collect relevant, timely research on my project topic, and synthesize it into a cohesive summary of existing knowledge in the field. From doing so, this project can come up with its own argument on this topic by conducting my own original research.

There are numerous research papers and journals that are related to this project's objective(s), specifically the various techniques used to create a pothole detection system, machine learning algorithms comparisons on pothole detection and the different results and outcome using varying deep learning methods.

### 2.2 Facts and Findings

According to The Federal road accidents caused by potholes have resulted in more than 5,000 deaths in India in the three years between 2018 and 2020, according to official data. As per the Ministry of Road Transport and Highways (MoRTH) data, the total number of road accident deaths due to potholes in 2018, 2019, and 2020 stood at 2,015, 2,140, and 1,471, respectively, totalling to 5,626 deaths. As many as 4,775 and 3,564 accidents have occurred in 2019 and 2020, respectively, due to potholes.

As for the statistics in Malaysia, in a study conducted by (Yusria Darma *et al*, 2017) revealed that uneven roads are amongst the many factors leading to road traffic fatalities, with 11.25% of which being associated with this, and potholes constituting 11.2% in Malaysia.

Accident cases due to potholes or bad roads are rising and as many as 223 accidents caused by poor road conditions especially potholes in Selangor were recorded between 2018 and 2020. According to the Selangor Traffic and Investigation Department, Chief Superintendent Azman Shari'at, 148 cases resulted in death. Out of these fatalities, 116 cases involved motorcyclists. In this article, some major accidents that took place caused by potholes were stated as well.

According to an article from The Vibes.com, a total of 285 deaths or an average of 57 fatalities annually due to pothole incidents were recorded over a period of five years from 2014 to 2018. (Tamrin, 2021) stated in the article that data obtained from the annual statistical report by the Royal Malaysia Police has shown that deaths due to potholes in 2018 added up to a mere 0.88% of overall road fatalities nationwide. Figure 1 shows the statistical bar graph of accidents involving potholes in Malaysia and in 2018, there were 55 deaths, with 16 suffering serious injuries and 40 minor injuries and the highest death count due to pothole incidents was 101 (in 2015), or 1.5% of overall road deaths.

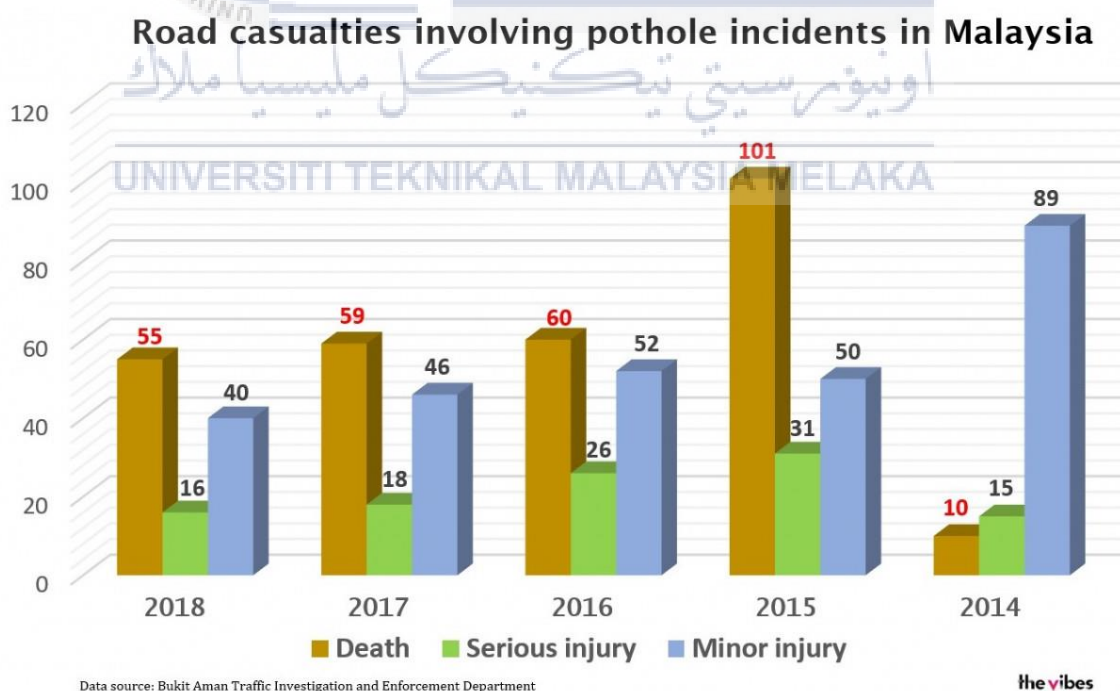


Figure 1: Statistics of accidents involving potholes in Malaysia.

### 2.2.1 Domain

One of the many domains of the pothole detection system, the very crucial one is computer vision which involves the development and implementation of algorithms and techniques to analyse visual data, such as images or videos, captured by cameras or sensors. Computer vision is mainly for object detection, image segmentation, feature extraction, and anomaly detection to identify and locate potholes.

Besides that, machine learning and deep learning domains are fundamental to pothole detection systems. These domains involve training models to recognize patterns in data and make predictions. Techniques like convolutional neural networks (CNNs), transfer learning, and data augmentation can be employed to train models for pothole detection. These domains are used for feature extraction, classification, and decision-making in pothole detection systems.

Pothole detection systems can utilize sensor networks embedded on vehicles to gather data related to pothole or road conditions. These sensors can capture information such as vibrations, accelerations, or changes in road surface characteristics, which can help in detecting and predicting the presence of potholes.

UNIVERSITI TEKNIKAL MALAYSIA MELAKA

Internet of Things (IoT) technologies enable the integration of various devices and sensors within the pothole detection system. This domain involves collecting data from connected devices, transmitting data wirelessly, and enabling communication between different components of the system.