



A

Project Report

On

# **Image Processing Based Semi-Automated Target Detector and Shooter**

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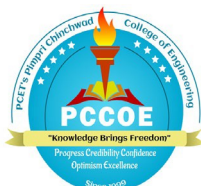
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(SAVITRIBAI PHULE PUNE UNIVERSITY, PUNE) 2020-21

# **CERTIFICATE**



## **Project Phase -II Report**

**On**

# **Image Processing Based Semi-Automated Target Detector and Shooter**

Submitted for Partial Fulfillment of the Requirements for the Degree of Bachelor  
of Engineering in the Department of Electronics & Telecommunication  
Engineering Pimpri Chinchwad College of Engineering, Savitribai Phule  
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## **LIST OF ABBREVIATIONS:**

DC – Direct Current

μC - Microcontroller

PCB - Printed Circuit Board

USB - Universal Serial Bus

CPU - Central Processing Unit

IC - Integrated Circuit

RPM - Revolutions Per Minute

DMM - Digital Multimeter

IDE - Integrated Development Environment

OS - Operating System

Li-ion - Lithium ion

ADC -Analog to Digital Convertor

GPIO - General Purpose Input Output

LAN - Local Area Network

Rpi – Raspberry Pi

YOLO – You Only Look Once

FPS – Frames Per Second

CNN – Convolution Neural Network

SPP - Spatial Pyramid Pooling

## **Acknowledgement**

All the accomplishments in the world require the effort of many people and this project is no different. Regardless of the source, we wish to express our gratitude to those who have contributed to the success of this project. We gratefully acknowledge and express heartfelt regards to all the people, who helped us in making the idea of the project a reality.

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## **Abstract**

With ever-growing increase in technology everything is getting automation. As we know currently India is not having good foreign relations with neighboring countries like China and Pakistan. These countries continuously trying to provoke war against us wherein we are losing our soldiers. Therefore, to reduce human casualties and to improve defensive systems at borderline we must upgrade our systems. Keeping this in mind this system is designed which emphasizes on fabrication of automated target detector and shooter robot.

The primary objective of this system is to serve a wireless automated machine which is better aesthetically. The system will be based on raspberry pi processor with pi camera and computer vision for its basic purpose and will be trained for specific object or event detection in real time video and will send the signal back to controller so that necessary actions will be taken. The appropriate processor is selected for the application so that it will handle the power requirement and processing capacity of all the necessary requirements. The greatest motivation behind this project is changing technology and greatest inspiration is Boston dynamics. This system with some advancement can be used for projectile target detection in hilly areas. By the end of the project one can use this system with very ease remotely from certain radio distance to control it and find a specified target with the help of system.

# ***CHAPTER :1***



# **Chapter 1**

## **Introduction**

### **1.1 Motivation**

India is a developing country. We as a country are always trying to push our limits by technological advancements. For sustaining in this era India has to bring developments in the technology as well as its self-defense capabilities. Self-defense is the important priority of our nation in 21<sup>st</sup> century. As we know currently India is not having good foreign relations with neighboring countries like China and Pakistan. These countries continuously trying to provoke war against us wherein we are losing our soldiers. Therefore, to reduce human casualties and to improve defensive systems at borderline we must upgrade our systems. The main motive of this project is to create a system which can replace humans at battlefields, which can be used remotely and which can help our nation to defend itself from threats of neighboring countries. Which is why we are designing a system which will be a prototype of real machine that can be deployed on borders to increase our defense capabilities and to reduce the casualties on the borders.

### **1.2 Background**

The kind of robots used widely are in industries, companies and in gaming appliances. But there are very few robots working on defense. Out there on borders all that needed is continuous lookout of any unusual movements happening nearby where very less robots are happen to be effective. But what if those robots are given vision and monitoring can be done more feasibly? This system is developed such that it can see the movements happening around and wirelessly from remote location and can work efficiently without any human casualties. We know that in near future all that human work is going to be replaced by the AI powered machineries. Taking this into consideration we came up with this idea of making computer vision-based robot which works automatically for detection of specific event and take actions accordingly.

### **1.3 Project Specification**

To design an integrated Computer Vision based system which captures real time video input and processes it to detect specific object, decides it as safe or threat and shoots at object.

## ***CHAPTER :2***

## Chapter 2

### Literature Survey

| Sr. No. | Title of the Paper   | Year of Publication | Publication/Conference   | Conclusion  |
|---------|--|---------------------|--|---|
| 1       | Digital image processing techniques – a survey   | 2016                | International Multidisciplinary Research Journal                     | Digital image processing deals with manipulation of digital images through a digital computer. In this paper various types of DIP technique presented in the literature are discussed and analysed. The DIP technique using image compression, edge detection and segmentation provides better compression ratio and accuracy of an image.  |
| 2       | Literature survey on the various methods of object detection in video surveillance systems | 2016                | International Research Journal of Engineering and Technology (IRJET) | In this paper, we have presented various studies, methods through which a better and clean picture of video surveillance can be projected. It is clearly stated that if all the steps of video analytics is taken and problem solving methods with its pros and cons are applied, an effective mechanism can be build up which will result in fruitful and clear image capturing. |
| 3       | Object detection with deep learning: a review  | 2019                | IEEE   | Due to its powerful learning ability and advantages in dealing with occlusion, scale transformation and background switches, deep learning-based object   |

|   |  |      |   |  |
|---|--|------|---|--|
|   |  |      |   | detection has been a research hotspot in recent years.   |
| 4 | Analytical description of pneumatic system                             | 2013 | International Journal of Scientific & Engineering Research        | We studied about various types of Pneumatic actuators and its working.   |
| 5 | Real-time object detection and tracking in an unknown environment      | 2011 | 2011 World Congress on Information and Communication Technologies | The proposed algorithm for object detection and tracking in unknown environment shall open new vista in field of computer vision for developing real world applications and also improvising currently existing algorithms to be operational in the real world.  |
| 6 | Real Time Object Detection and Tracking Using Deep Learning and OpenCV | 2018 | IEEE  | Deep learning has gained a tremendous influence on how the world is adapting to Artificial Intelligence since past few years. Some of the popular object detection algorithms are Region-based Convolutional Neural Networks (RCNN), FasterRCNN, Single Shot Detector (SSD) and You Only Look Once (YOLO). Amongst these, Faster-RCNN and SSD have better accuracy, while YOLO performs better when speed is given preference over accuracy. Deep learning combines SSD and Mobile Nets to perform efficient implementation of detection and tracking. This algorithm performs |

|   |   |      |      |   |
|---|---|------|------|---|
|   |   |      |      | efficient object detection while not compromising on the performance.   |
| 7 | Research of the Real-time Detection of Traffic Flow Based on OpenCV | 2008 | IEEE | <p>A vehicle detection algorithm was proposed based on the morphology and wavelet transform, in the context of the traditional difference. First, the background model was established, using statistical means of the rapid sequence. As background to transform the impact of light obviously, the corresponding easy and quick to update the background algorithm was used. Using the background of the video images to do background subtraction, and then images of the vehicles were accurate detection of mathematical morphology and wavelet transform. A video vehicle detection system was developed using visual C++6.0 and OpenCV image and development kits. A highway traffic flow has been detected by a background extraction, image filtering, image binary, morphological transformation, vehicle detection and segmentation methods and steps. To achieve some highway traffic flow analysis, results showed that: the system to identify the correct rate of more than 98 percent, satisfying the requirements of practical applications.</p> |

|   |   |      |                    |   |
|---|---|------|--------------------|---|
| 8 | Robust Real-time Object Detection                       | 2015 | Cambridge Research | <p>This paper describes a visual object detection framework that is capable of processing images extremely rapidly while achieving high detection rates. There are three key contributions. The first is the introduction of a new image representation called the “Integral Image” which allows the features used by our detector to be computed very quickly. The second is a learning algorithm, based on AdaBoost, which selects a small number of critical visual features and yields extremely efficient classifiers</p>  |
| 9 | You Only Look Once: Unified, Real-Time Object Detection | 2016 | IEEE               | <p>Prior work on object detection repurposes classifiers to perform detection. Instead, we frame object detection as a regression problem to spatially separated bounding boxes and associated class probabilities. A single neural network predicts bounding boxes and class probabilities directly from full images in one evaluation. Since the whole detection pipeline is a single network, it can be optimized end-to-end directly on detection performance. Our unified architecture is extremely fast. Our base YOLO model processes images in real-time at 45 frames per second. A smaller version of the network, Fast YOLO, processes an</p> |

|    |   |      |      |  |  |
|----|---|------|------|--|--|
|    |   |      |      | astounding 155 frames per second while still achieving double the mAP of other real-time detectors. Compared to state-of-the-art detection systems, YOLO makes more localization errors but is less likely to predict false positives on background. Finally, YOLO learns very general representations of objects. It outperforms other detection methods, including DPM and R-CNN, when generalizing from natural images to other domains like artwork.   |  |
| 10 | Low-cost smart security camera with night vision capability using Raspberry Pi and OpenCV | 2014 | IEEE | In order to further maintain peace and provide security to people now a day, Closed-circuit television (CCTV) surveillance system is being utilized. This study focused on the design and implementation of a low-cost smart security camera with night vision capability using Raspberry Pi (RPI) and OpenCV. The system was designed to be used inside a warehouse facility. It has human detection and smoke detection capability that can provide precaution to potential crimes and potential fire. The credit card size Raspberry Pi (RPI) with Open-Source Computer Vision (OpenCV) software handles the image processing, control algorithms for the alarms and sends captured pictures to user's email via Wi-Fi. |  |

|    |  |      |  |  |
|----|--|------|--|--|
| 11 | Study on Object Detection using Open CV – Python   | 2017 | International Journal of Computer Applications (0975 – 8887)   | It is concluded that comparing images using colour, texture, and shape are not enough because two objects might have same attributes.  |
| 12 | Vision-andLidar Based Real-time Outdoor Localization for Unmanned Vehicles                   | 2018 | IEEE International Conference on Information and Automation Wuyi Mountain, China                           | A real-time, effective onboard outdoor localization system with a depth camera and a laser scanner shows the maximal utility of image information and laser information to improve the robustness and accuracy of pose estimation. |
| 13 | Design, Realization and Sensorization of the Dexterous iCub Hand                             | 2010 | International Conference on Humanoid Robots Nashville, TN, USA, December 6-8, 2010                         | The hand is the result of a design that optimized the level of integration of the hand in the overall robot to meet the project specifications in terms of dimensions, dexterity and sensorization.                                |
| 14 | Design of Intelligent Mobile Robot System Based on Ultrasonic Sensors                        | 2007 | Design of Intelligent Mobile Robot System Design of Intelligent Mobile Robot System Automation and Systems | By using the design implemented for this paper, the sensor pods could be integrated with other mobile robots to provide noncontact sensing and navigation for them as well.  |
| 15 | Image processing and machine learning techniques used in computer-aided detection system for | 2020 | International Journal of Electrical and  | In this paper we came to know that neither a single technique is applicable to all types of images nor all the techniques perform well for one particular image. Furthermore, none of  |



|  |                                  |  |                                    |   |
|--|----------------------------------|--|------------------------------------|---|
|  | mammogram screening-<br>A review |  | Computer<br>Engineering<br>(IJECE) | the segmentation procedure is fully<br>automatic. So, machine learning based<br>intelligent systems can help to make the<br>complete procedure automated. |
|--|----------------------------------|--|------------------------------------|---|

**Literature Survey Summary:**

These papers gave us idea about latest technological trends not only in electronics but also from mechanical background which have major contribution in the robotics field. Most of the focus of image processing-based computer vision projects is to implement algorithms such as reinforcement learning, archer algorithm, neural network algorithms in order to provide better efficiency to its applications. These papers gave a brief idea about how an image can be manipulated and processed in order to get effective outcome such as improving the surveillance camera captures, etc. Besides electronics we have also studied how Mechanical systems are used in robot actuations and various movements. We have discovered how a pneumatic system can be used in our project based on its operation or mode of actuation to give a proper thrust or force to our arrow.

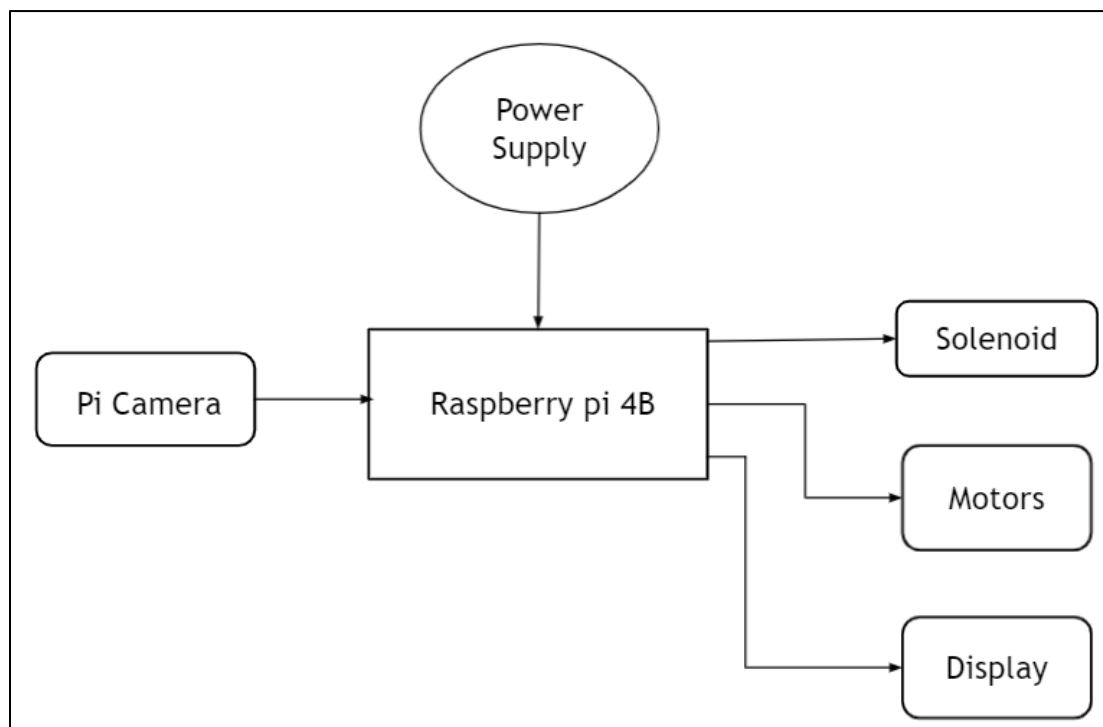
## ***CHAPTER :3***

## Chapter 3

### Methodology

The implementation of the project using a block structure is explained here, it shows the hardware components used, with block wise flow according to which the project works.

#### 3.1 Block Diagram:



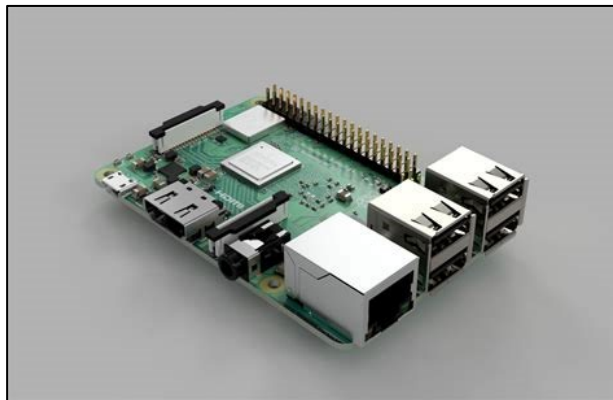
**Fig. 3.1** Block Diagram

### 3.1.1 Block Diagram Explanation:

The proposed system works as a normal machine does. It takes several inputs and gives the required output. The main central processing unit of robot is Raspberry Pi 4 model B. The inputs to raspberry pi are Pi camera, Power supply and trained weights from the memory card. The pi camera is used for real time image and video capture and the trained weights used are generated using machine learning which will help to identifying the target. Power supply is connected in order to work of real time image/video processing inside the controller. The power supply provided is using Li-ion batteries. The robot after processing the input produces several outputs too. After processing the video this video is then transferred to the display for user interaction. Another output is in terms of motor actuation which is used for robot to align itself such that it aligns with the target. After aligning itself with target and having proper actuation with help of push pull solenoid to shoot target.

### 3.2 Elements of Block Diagram:

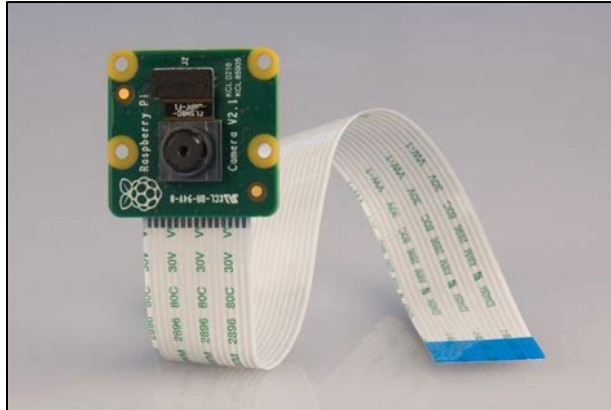
#### 3.2.1 Raspberry Pi 4 Model B:



**Fig 3.2.1** Raspberry Pi 4 Model B

The Raspberry Pi 4 Model B is advanced version of Raspberry Pi 4 series. It is working on 64-bit quad core processor with the clock frequency of 1.44Ghz and 2 GB ram. It is basically the CPU of our robot which will carry out all the necessary actions required for object detection and motor actuation. It also supports dual band 2.4GHz and 5GHz which will be used for signal transfer to and from the Raspberry Pi to user interface such as signaling motor actuation and retrieving the live camera footage. It is also provided with in-built Bluetooth 4.2/BLE module which can be used as alternative to Wi-Fi communication. The I/O port of Raspberry Pi is used for interfacing the Pi camera which is compatible with Raspberry Pi. For memory storage purpose it has been given Micro SD port in which OS and necessary code and weights can be stored.

### 3.2.2 Pi camera:



**Fig 3.2.2** Pi Camera

To identify the target using image processing algorithm the basic requirement is an image. So, the Pi camera is used for providing the source of images to the Raspberry Pi. Pi camera is compatible with Raspberry Pi and has a 5MP camera for better image capture.

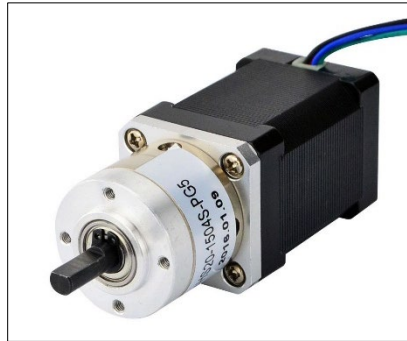
### 3.2.3 ADC0804 IC:



**Fig 3.2.3** ADC0804 IC

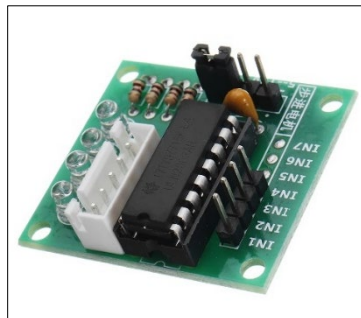
To convert the sensor signal/analog signal into the digital one to transmit it between robot and user interface/display wirelessly it is required for the data to be in digital form. So, the ADC804 is used which is 8 pin IC which is used to convert the analog signal to digital signal.

### 3.2.4 Motor and Motor drivers:



**Fig 3.2.4.a** Stepper Motor

A stepper motor acts in steps and hence can provide very precise movement. Also, the motors position can be commanded and hold at one step using the magnetic locking mechanism which provide high torque for stability purpose. To carry out the vertical movement of robot body to align itself towards the target stepper motor is used.



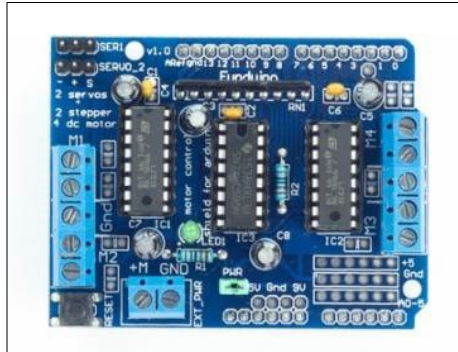
**Fig 3.2.4.b** ULN2003

ULN20003 is a motor driver IC, which is used to drive a stepper motor to take a precise step angular movement.



**Fig 3.2.4.c** DC Motor

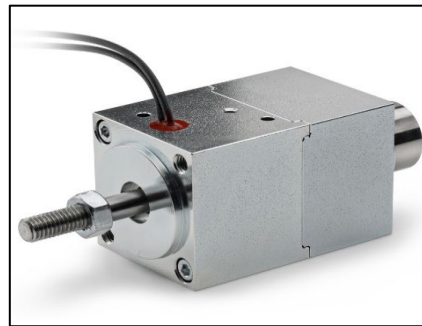
Simple DC motors which convert electrical energy into mechanical energy are used to drive the robot. These motors are taken in application considering the power requirement by the motors and the power source used in the project. The power source used is sufficient to drive DC motors. 4 DC motors are used as 4 different wheels of the robot which will be having 4-wheel drive mechanism.



**Fig 3.2.4.d L293D IC**

L293D IC is a DC motor driver IC which is capable of driving 2 DC motors at a time. There are times which the robot has to move front and back and sometimes no movement. L293D is used to drive the motors in bidirectional mode i.e., clockwise and anticlockwise using a direct current.

### 3.2.5 Push-Pull Solenoid:



**Fig 3.2.5: Push - Pull Solenoid**

Fig 3.2.5 shows a Push – Pull Solenoid which is an Electro-mechanical device which is used to hit the shooting material giving an impulse to it in the forward direction to hit the target.



### 3.2.6 Power supply:



**Fig 3.2.6** Li-ion Batteries

Currently widely used batteries are lithium-ion battery or Li-ion battery. These batteries are commonly used because of its low density and better efficiency as compared to Lead acid batteries which are heavy and more hazardous to nature. Li-ion batteries are used as the base power supply for the robot system and Raspberry Pi controller.

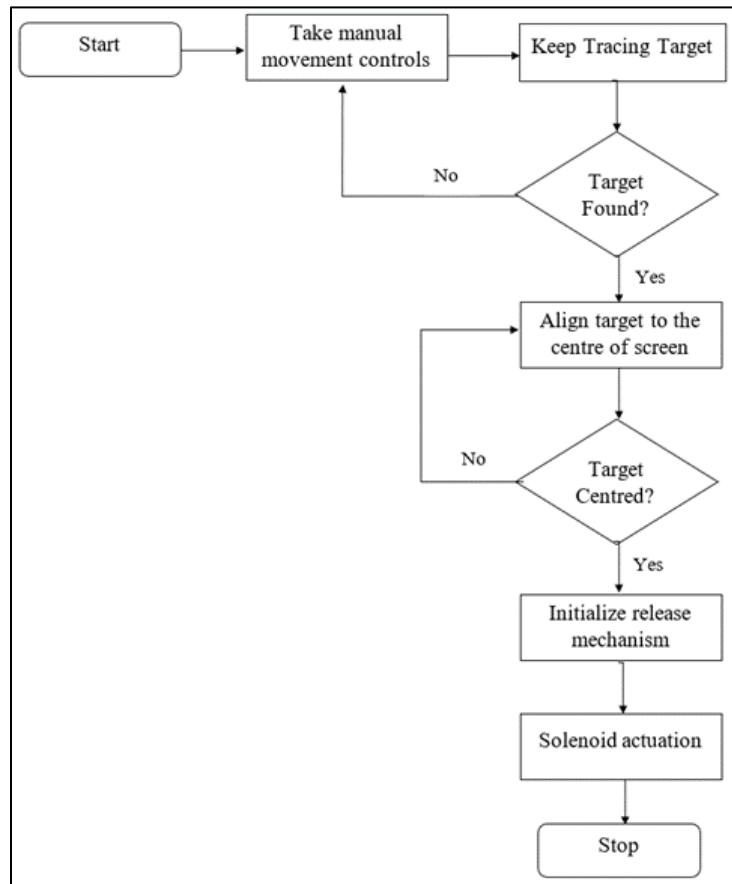
### 3.2.7 Display:



**Fig 3.2.7** Display

A mobile phone display to see what the camera is capturing and to see if the target is properly in the frame for further process. Also, this display is provided with the controls which will operate the robot remotely and will be responsible for movements of robot.

### 3.3 Flow Chart:



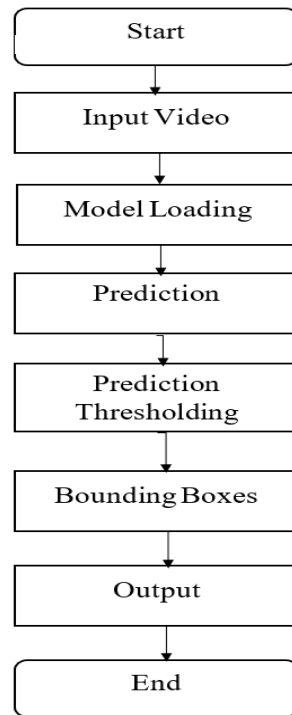
**Fig 3.3.1** System Flow Chart

#### 3.3.1 Flow Chart Explanation:

The robot starts taking manual control action and starts tracing the target in camera as soon as it turns on. The robot keeps on searching the target until it detects target into the camera screen, as soon as the target is found anywhere in pi camera the robot performs following operations in order.

1. Identify the target
2. Aligning robot such that the target is centred.
3. Give feedback to controller and wait for further instructions.

When controller confirms the target and releases shoot command then the power is applied to the shooter through pneumatic actuator and target is been hit.



**Fig 3.3.2** Target Detection Flowchart

As per the Fig 3.3.2 as soon as system starts tracing target it will take input video for processing via Pi Camera.

- It will then load the model which we have trained to detect the particular target. These model loading consists of weights and cfg(configuration) file.
- After model loading system starts prediction. It starts to predict whether the input video has the target/object which is to be traced.
- There is a prediction thresholding which should be set so that if the prediction rate crosses the threshold, then system can be sure if that is the required target.
- After crossing threshold value, it creates bounding boxes around the target with certain amount of confidence level so that target will be easily visible on the display.
- These bounding boxes will act as an output which will be provided to the user. This real time processing of video makes it easier to identifying potential threats.

## ***CHAPTER :4***

## Chapter 4

### Software Implementation

#### 4.1 Anaconda IDE:



**Fig 4.1** Anaconda IDE

Conda works on command line interface or anaconda prompt on windows which uses command line interface. To have various testing conda offers variety of packages. Basically, anaconda IDE is used as an environment or development tool for python program applications. Since our programming language is python, we opted for this as an IDE. Also, by using simple commands in anaconda prompt various testing of pi camera is done. Before actually going into real time processing the basic testing on image processing algorithm is done in Anaconda using python programming and OpenCV and numpy libraries.

#### 4.2 Google Colab:



**Fig 4.2** Google Colab

Google colab or google collaborator is an online product which is made available by google research. Using colab one can write and execute the arbitrary python code on google server through browser. Google colab provides a virtual platform for development which comprises of high-speed processors with clock frequency in terms of TB's. We personally used this platform to train the supervised machine learning model using neural networks and labelled images. We have implemented the machine learning model by using transfer learning which used an algorithm YOLO (You only look once) which requires a image data in labelled form. We used this algorithm because it is proven to be the effective real-time object recognition algorithm.

### 4.3 YOLOv4 Algorithm:

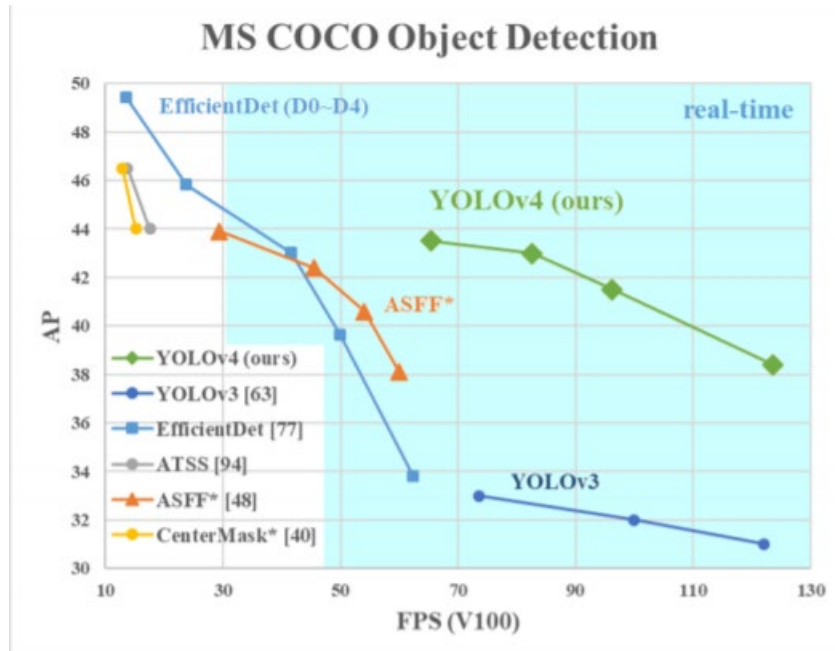


Fig 4.3.1 MS COCO Object Detection [13]

Today's World has rapidly developing environment and needs fast processing at smooth and faster rate. The above graph shows the comparison of YOLOv4 algorithm with its competitive algorithms used for real-time object detection. Here we can clearly see that YOLOv4 is always a better option wherein it is able to process a greater number of frames (FPS) with greater mean average Precision (AP). There are some algorithms which have greater AP but can process a very smaller number of frames at a time which is not feasible in great automation sectors. So as precision and speed both are essential parts of the project the YOLOv4 algorithm is used. Also, it is better than its previous version where the AP is increased by 10% and the FPS is increased by 12%.

- **One Stage Detection Model:**

Basic two stage detection models predict bounding boxes for each object and for each object found it then classifies the probabilities of each object so this way the model speed gets reduced wherein in one stage detection model, it requires only a single pass to predict all the bounding boxes as well as to classify the probability of each object. As YOLOv4 is using the one stage detection modelling for object detection it is a faster way to get the object bounding boxes and its probabilities in real time scenarios.

- **Architecture:**

YOLOv4 is consisting of following parts:

1. Backbone:

In YOLOv4 Backbone is referred as feature-extraction architecture in which according to the different name of YOLO, different backbones are used. In this project Darknet52 backbone is used for object detection which means that there are 53 layers of convolutional layers in

between input and the successful output for better accuracy. It is bit slower to train than other backbones such as YOLO-tiny which uses 9 layers of CNN but here accuracy is major concern so darknet53 has been chosen.

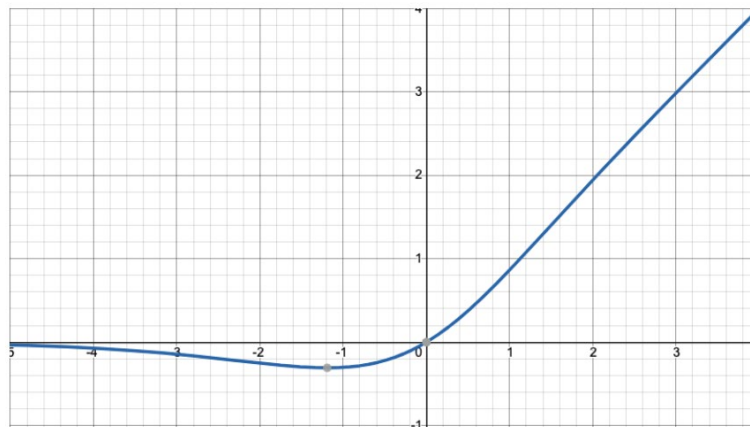
## 2. Neck:

To detect the object from different scales and different resolutions, a hierarchical structure is produced in between backbone and head to have feature mapping at different resolutions.

This part also consists several designs. During training of data, the network requires fixed size for each data but we don't have fixed sized data all the time. Spatial pyramid pooling (SPP) is one of the layers that removes the fixed sized constraint. SPP layer performs the information aggregation to avoid the need of cropping or wrapping or resizing of image.

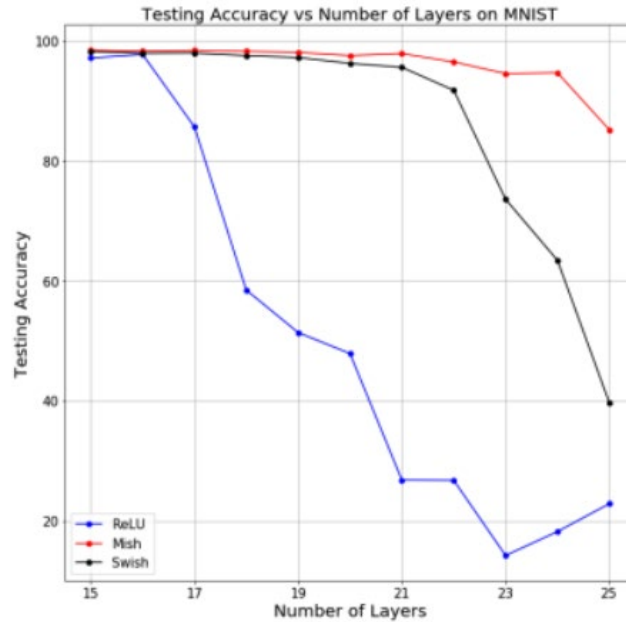
- **Activation Function:**

There are various activations functions which can be used such as ReLU, leaky-ReLU, parametric-ReLU, ReLU6, SELU, Swish, or Mish. Following is the Mish Activation function.



**Fig 4.3.2** Mish Activation Function [12]

The mish function used in YOLOv4 because of its low cost and various properties which improves the model's performance as compared to other activation functions. The non-monotonic property of Mish helps to preserve small negative values thereby stabilizing gradient flow.



**Fig 4.3.3** Accuracy Comparison Graph [12]

As shown in above graph the Mish activation function gives better testing accuracy at higher number of hidden layers so it is preferred activation function for the Darknet53 architecture.

The equation of Mish function is given by,

$$f(x) = x \cdot \tanh(\zeta(x))$$

where,  $\zeta(x) = \ln(1+e^x)$ , is a SoftMax activation function.



#### 4.4 YOLOv4-tiny Algorithm:

In proposed system for video processing, we used YOLOv4 as well as YOLOv4-tiny algorithm to find out which one is better at processing real time video. Here are some key features of YOLOv4-tiny.

- ▶ One stage Detection Model
- ▶ Compressed Version of YOLOv4
- ▶ Consists of 29 pretrained CNN layered architecture
- ▶ Neck (SPP) removes the fixed size constraint
- ▶ 10% more mean AP than YOLOv3
- ▶ 8 times FPS Support than YOLOv4
- ▶ 5/2 times FPS Support for proposed model
- ▶ Computational Complexity CSPBlock to ResBlock-D is 10:1
- ▶ Complexity ratio for proposed model is 4:1
- ▶ Train 350 images in 1 hour when using a Tesla P100 GPU
- ▶ For our model - approx. 1 hour and for Yolov4 model - 3-4 hours (6000 iterations) Tesla T4 CPU
- ▶ YOLOv4-tiny has an inference speed of 3 ms on the Tesla P100, making it one of the fastest object detection models to exist.

| <b>YOLOv4</b>                   | <b>YOLOv4-tiny</b>               |
|---------------------------------|----------------------------------|
| Works better with images.       | Works better with videos.        |
| Low FPS support (upto 55 FPS).  | High FPS support (upto 155 FPS). |
| More time complexity.           | Less time complexity.            |
| More space complexity.          | Less space complexity.           |
| Training average loss: 0.314525 | Training average loss: 0.012734  |

**Table 4.4:** YOLOv4 vs YOLOv4-tiny

## ***CHAPTER :5***

## Chapter 5

### Testing and Troubleshooting

#### 5.1 Testing:

The figure 5.1.1 shows the testing of pi camera module.



Fig 5.1.1 Testing of Pi camera module

## 5.2 Testing Strategies and Test Procedures:

### 5.2.1 Object Detector Model Training:

For testing purpose of machine learning model, we have provided a set of 58 different images to the ML model. The system or algorithm used by us is YOLOv4-tiny. YOLO stands for “You only look once”. We chose this algorithm for our project because latest Yolo system can process 45 frames per second and can enhance the real time detection system and since we are working on real time video processing, we chose this algorithm. The model was trained to extract the features and store the weights of given data images. We then tested the accuracy of our project and we found it to be working quite good with small set of images only. The following images are tested by using Yolo algorithm.

You can find the code here:

<https://drive.google.com/file/d/17rsfDBS1aiwTL6Yr9gBpm8znk-Wl7lMq/view?usp=sharing>

## 5.3 Results:

### 5.3.1 Object Detector Model Training:



Fig 5.3.1 Object Detector Model Training



### 5.3.2 Comparison Based on Simulations:

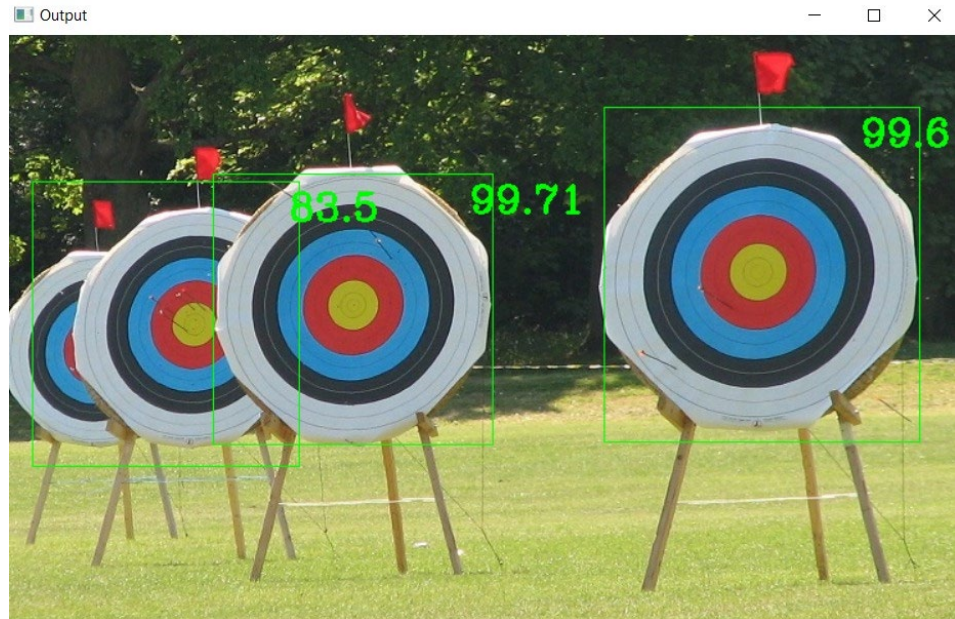


Fig 5.3.2.a: YOLOv4

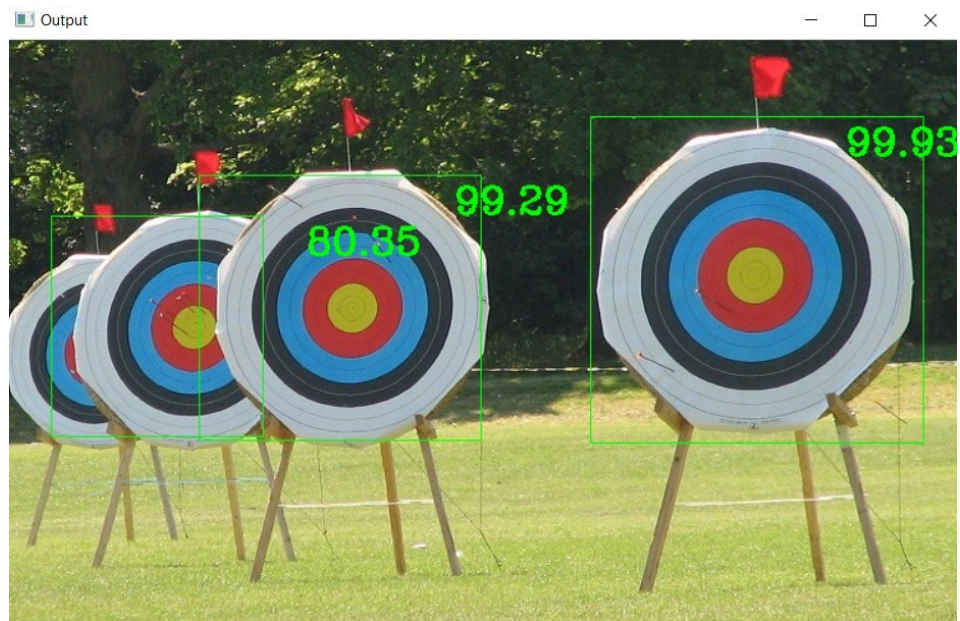


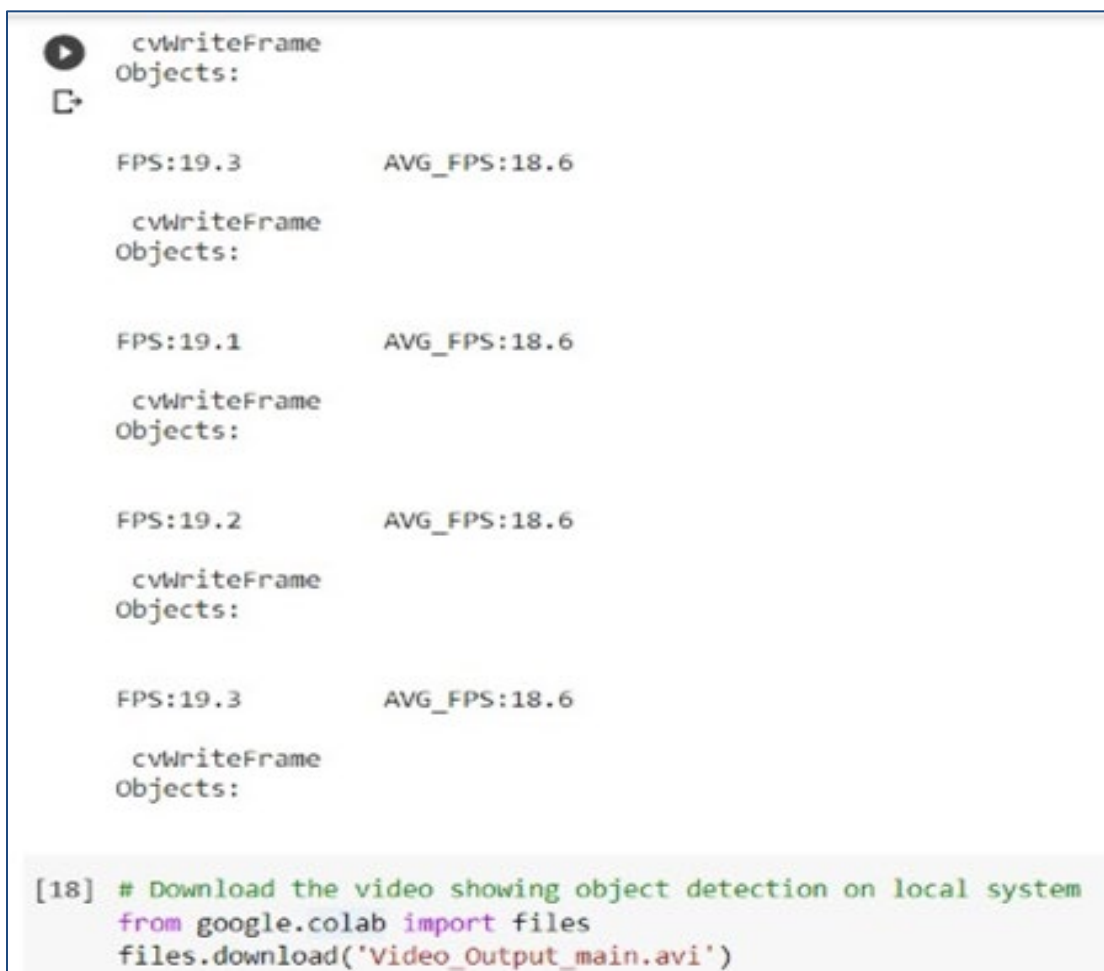
Fig 5.3.2.b: YOLOv4-tiny

### Comparison Results:

- YOLOv4 Average Precision: 94.27
- YOLOv4-Tiny Average Precision: 93.90
- Impact on average precision: ~0.5%

Here we can see that for Image processing YOLOv4 algorithm has provided average confidence level of 94.27 whereas YOLOv4-tiny algorithm has provided average confidence level of 93.90. Hence, we can say that YOLOv4 have better precision than YOLOv4-tiny for image processing.

Following are the Frames per second (FPS) results obtained from simulation:



```

cvWriteFrame
Objects:

FPS:19.3      AVG_FPS:18.6

cvWriteFrame
Objects:

FPS:19.1      AVG_FPS:18.6

cvWriteFrame
Objects:

FPS:19.2      AVG_FPS:18.6

cvWriteFrame
Objects:

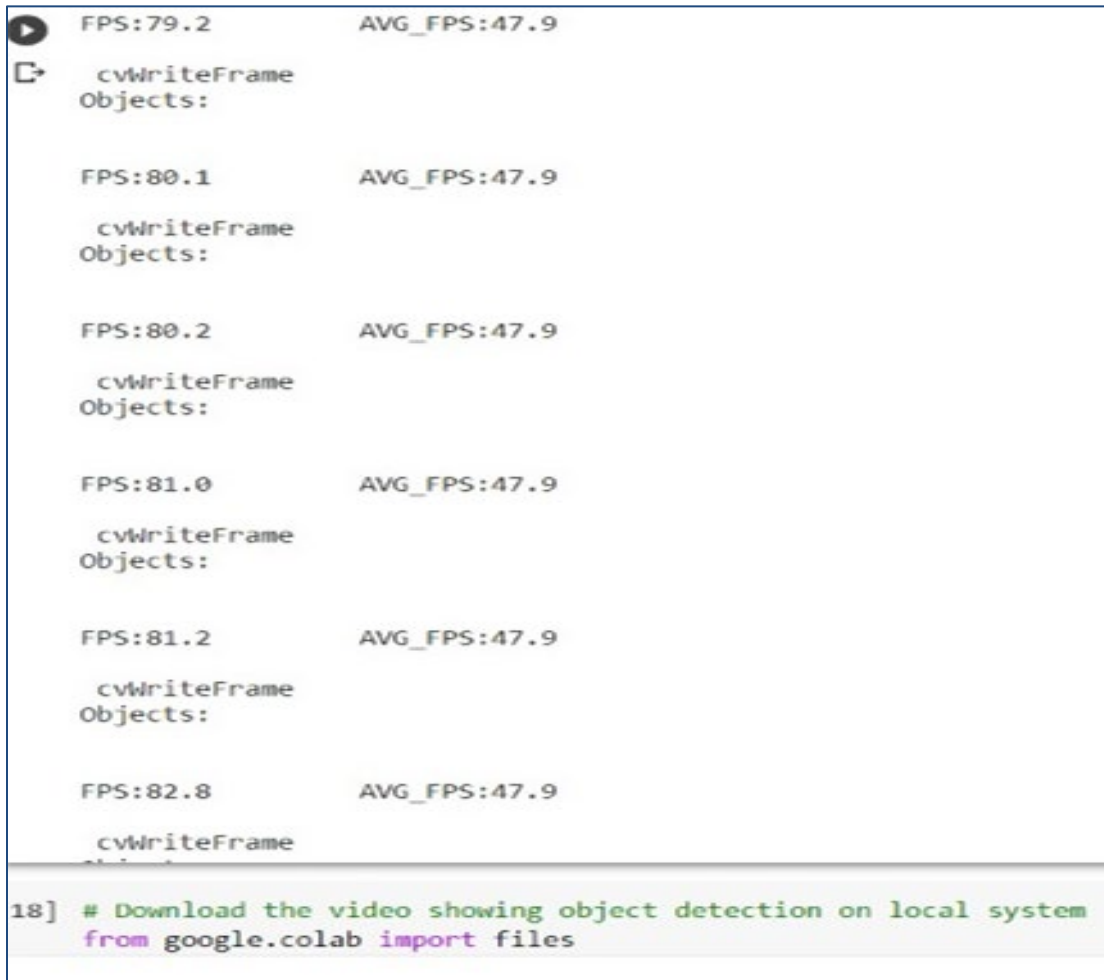
FPS:19.3      AVG_FPS:18.6

cvWriteFrame
Objects:

[18] # Download the video showing object detection on local system
from google.colab import files
files.download('Video_Output_main.avi')

```

**Fig 5.3.2.c: YOLOv4 FPS**



```

FPS:79.2      AVG_FPS:47.9
cvWriteFrame
Objects:

FPS:80.1      AVG_FPS:47.9
cvWriteFrame
Objects:

FPS:80.2      AVG_FPS:47.9
cvWriteFrame
Objects:

FPS:81.0      AVG_FPS:47.9
cvWriteFrame
Objects:

FPS:81.2      AVG_FPS:47.9
cvWriteFrame
Objects:

FPS:82.8      AVG_FPS:47.9
cvWriteFrame
Objects:

18] # Download the video showing object detection on local system
    from google.colab import files

```

**Fig 5.3.2.d: YOLOv4-tiny FPS**

From fig 5.3.2.c and fig 5.3.2.d we can clearly observe that for Video processing YOLOv4-tiny is better as compared to YOLOv4. We can see that YOLOv4 algorithm process average FPS of 18.6 while on the other hand YOLOv4-tiny algorithm provides processing in average FPS of 47.9 which is 3x more than YOLOv4 algorithm. Hence, we can say that YOLOv4-tiny is better for Video processing than YOLOv4 algorithm.



## ***CHAPTER :6***

## **Chapter 6**

### **Advantages and Applications**

#### **6.1 Advantages**

1. System can replace Humans on the battlefield and therefore helps in reducing the human casualties on borders.
2. Detection is improved due to real time image processing capabilities.
3. Accuracy of hitting the target is increased.
4. Human error can be reduced due to human-machine interface.
5. System can be controlled over long distance.

#### **6.2 Applications**

1. Military application i.e., target detection and shooting purpose.
2. Can be used as a Tranquilizer gun for animals in sanctuaries to inject anesthetic drug or tranquilizer.

## ***CHAPTER :7***

## **Chapter 7**

### **Conclusion**

The project plans to propose a system which would detect specific target objects with the help of video processing algorithm. Once the target object is detected, system controller shoots aiming at the same targeted object. Maximum accuracy is achieved by using pneumatic based subsystem. This proposed system consisting of wireless communication provides long range contactless operability to the users. Hence, this system in future can replace human- beings and therefore help in reducing human casualties on the border.

## References

- [1] <https://github.com/tzutalin/labelImg>
- [2] <https://github.com/AlexeyAB/darknet>
- [3] P.F. Felzenszwalb and D. P. Huttenlocher, 2005. Pictorial structures for object recognition. IJCV, 61(1):55–79, 2005.
- [4] Y. Ren, C-S. Chua and Y-K. Ho, 2003. Motion Detection with Nonstationary Background, MVA, Springer-Verlag.
- [5] P. F. Felzenszwalb, R. B. Girshick, D. McAlester, and D. Ramanan, “Object detection with discriminatively trained part-based models,” IEEE Trans. Pattern Anal. Mach. Intell., vol. 32, no. 9, p. 1627, 2010.
- [6] P. L. Rosin, “A simple method for detecting salient regions,” Pattern Recognition, vol. 42, no. 11, pp. 2363–2371, 2009.
- [7] C. Szegedy, A. Toshev, and D. Erhan, “Deep neural networks for object detection,” in NIPS, 2013.
- [8] B Gupta, Ashish Chaube ,” Study on Object Detection using Open CV – Python,2017”, International conference of Computer applications, doi: 10.5120/ijca2017913391
- [9] Chen, Haipeng & He, Zhentao & Shi, Bowen & Zhong, Tie. (2019). Research on Recognition Method of Electrical Components Based on YOLO V3. IEEE Access. 7. 1-1. 10.1109/ACCESS.2019.2950053.
- [10] Cancan Yang, Zan Yang, Shan Liao, Zheming Hong, Wei Nai.(2020) Triple-GAN with Variable Fractional Order Gradient Descent Method and Mish Activation Function.

- [11] A. M. Ghoreyshi, A. AkhavanPour and A. Bossaghzadeh, "Simultaneous Vehicle Detection and Classification Model based on Deep YOLO Networks," 2020 International Conference on Machine Vision and Image Processing (MVIP), 2020, pp. 1-6, doi: 10.1109/MVIP49855.2020.9116922.
- [12] Z. Zhang, Z. Yang, Y. Sun, Y. Wu and Y. Xing, "Lenet-5 Convolution Neural Network with Mish Activation Function and Fixed Memory Step Gradient Descent Method," 2019 16th International Computer Conference on Wavelet Active Media Technology and Information Processing, 2019, pp. 196-199, doi: 10.1109/ICCWAMTIP47768.2019.9067661.
- [13] J. Redmon, S. Divvala, R. Girshick and A. Farhadi, "You Only Look Once: Unified, Real-Time Object Detection," 2016 IEEE Conference on Computer Vision and Pattern Recognition (CVPR), 2016, pp. 779-788, doi: 10.1109/CVPR.2016.91.
- [14] J. Huang, Y. Lu, "A Method for Identifying and Classifying Resistor and Capacitors Based on YOLO Network," 2019 IEEE 4th International Conference on Signal and Image Processing, 2019.