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Department of Electronics & Telecommunication Engineering

PROJECT PHASE 2 SEMINAR

“TRAFFIC RULES VIOLATION RECOGNITION FOR TWO-WHEELER USING DEEP LEARNING”

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Contents :

- Aim of the project
- Introduction
- Block diagram
- Objectives of the Project
- Design
- Software implementation
- Results
- Conclusion
- Future Scope
- Advantages
- Applications
- References

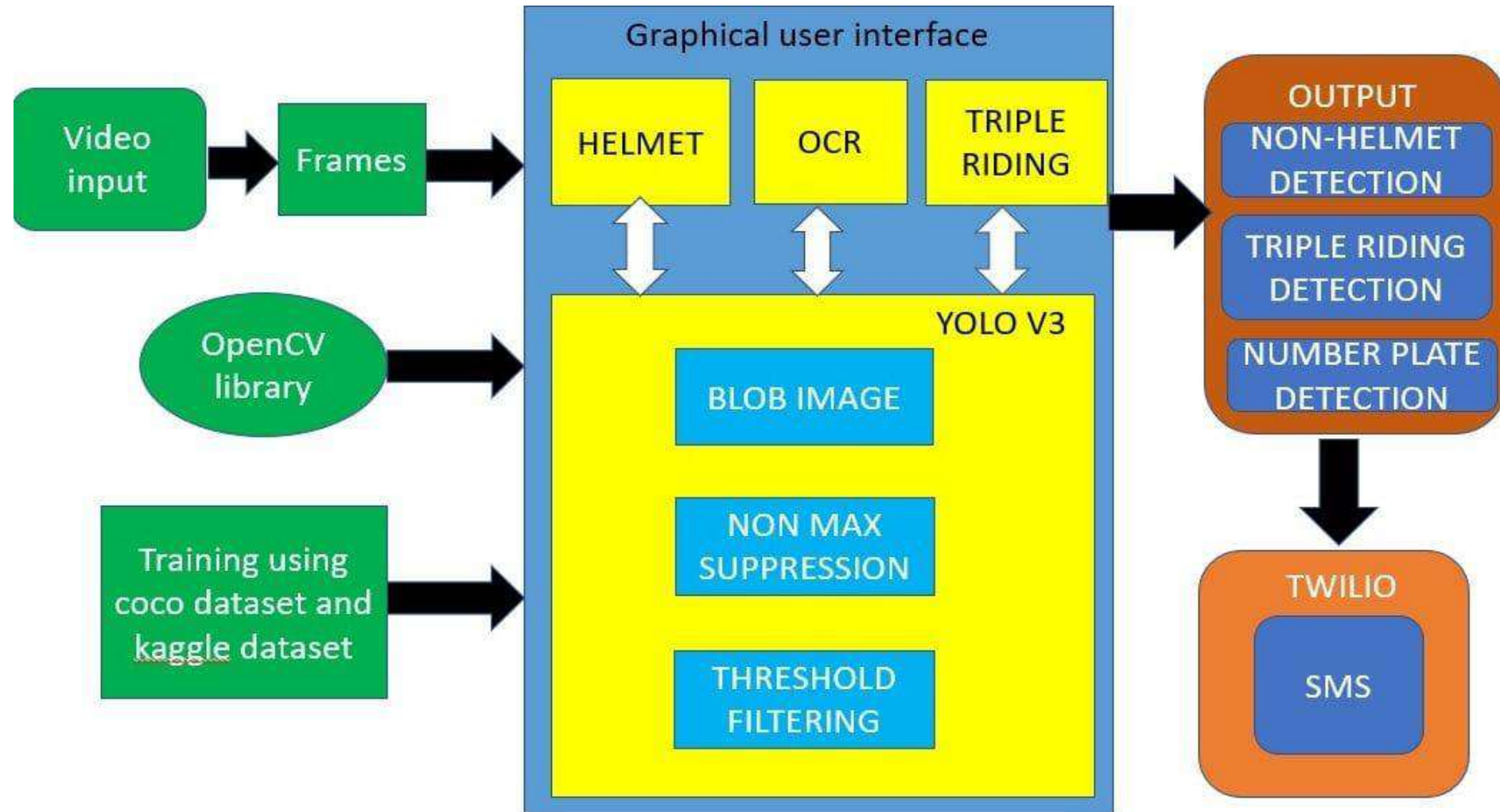
Aim of the project :

- We usually prefer motorbikes over other vehicles as it is significantly less expensive to run, less demanding to park and adaptable in rush hour gridlock.
- In India, in excess of 118 million individuals are utilizing bikes. Since lot of bikes travel along with other vehicles wearing headgear is critical to decrease the danger of injuries.
- we propose an approach where we identify the motorcycle riders without headgear, who are triple riding and who are using mobile phones while driving by utilizing surveillance videos in real-time.
- Our approach also proposes a system where a message will be sent to the concerned authority about the vehicle details of the person who violates the above-mentioned rules with the proof.

INTRODUCTION :

- Bike is an extremely mainstream method of transportation in India. However, there is a high risk involved due to lack of protection.
- To decrease the involved risk, it is highly desirable for motorcycle riders to use helmet. That's why the government has made it a punishable offense to ride a bike without helmet.
- The drawback of the current method where human intervention is required can be solved by our proposed method.
- Nowadays everyone is moving towards automation and many countries have implemented automatic traffic surveillance system.
- Here we are developing a system in which we are using a more efficient way which helps us in getting better results.

Block Diagram :



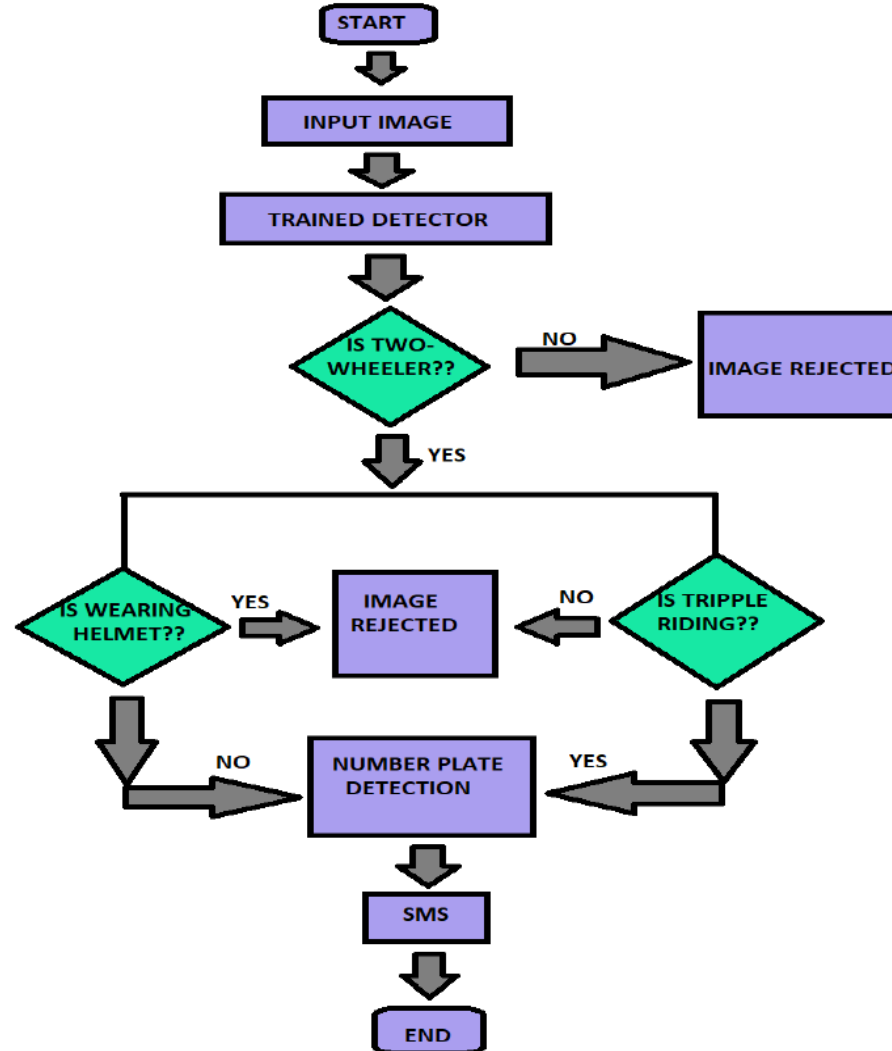
Objectives :

The main objective of the project work is to develop a system for

- Detection of persons on the motorbikes who are not wearing the helmet and triple riding while crossing the signals and junctions .
- Sending an SMS to the concerned authority about the vehicle details of the person who violates the above-mentioned rule.
- To improvise the metric(accuracy,speed and efficiency) to yield better results

Design :

Flow Diagram



STEP1 - A real time input video is fed to the model.

STEP2 - The model is previously trained with a dataset. Then the real time input is compared with dataset.

STEP3 - If the two-wheeler is detected in the input the process continues else the image is discarded.

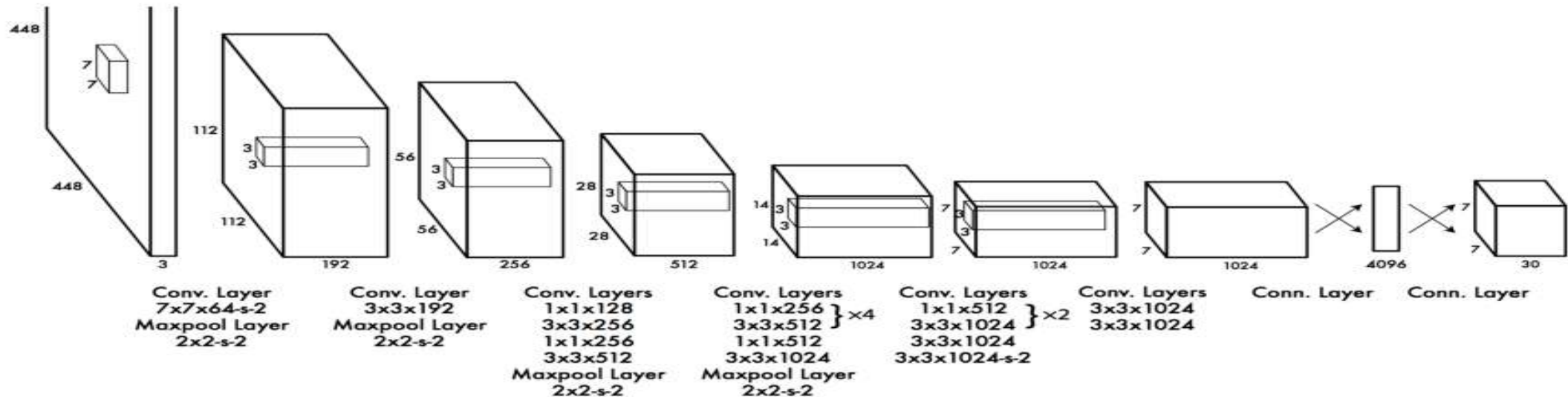
STEP4 - After two-wheeler is detected, the pretrained model detects the rider wearing helmet and triple riding. If the rules are violated it goes to the next step else the image is discarded.

STEP5 - When the rules are violated an SMS alert will be sent to the concerned authority.

Algorithm :

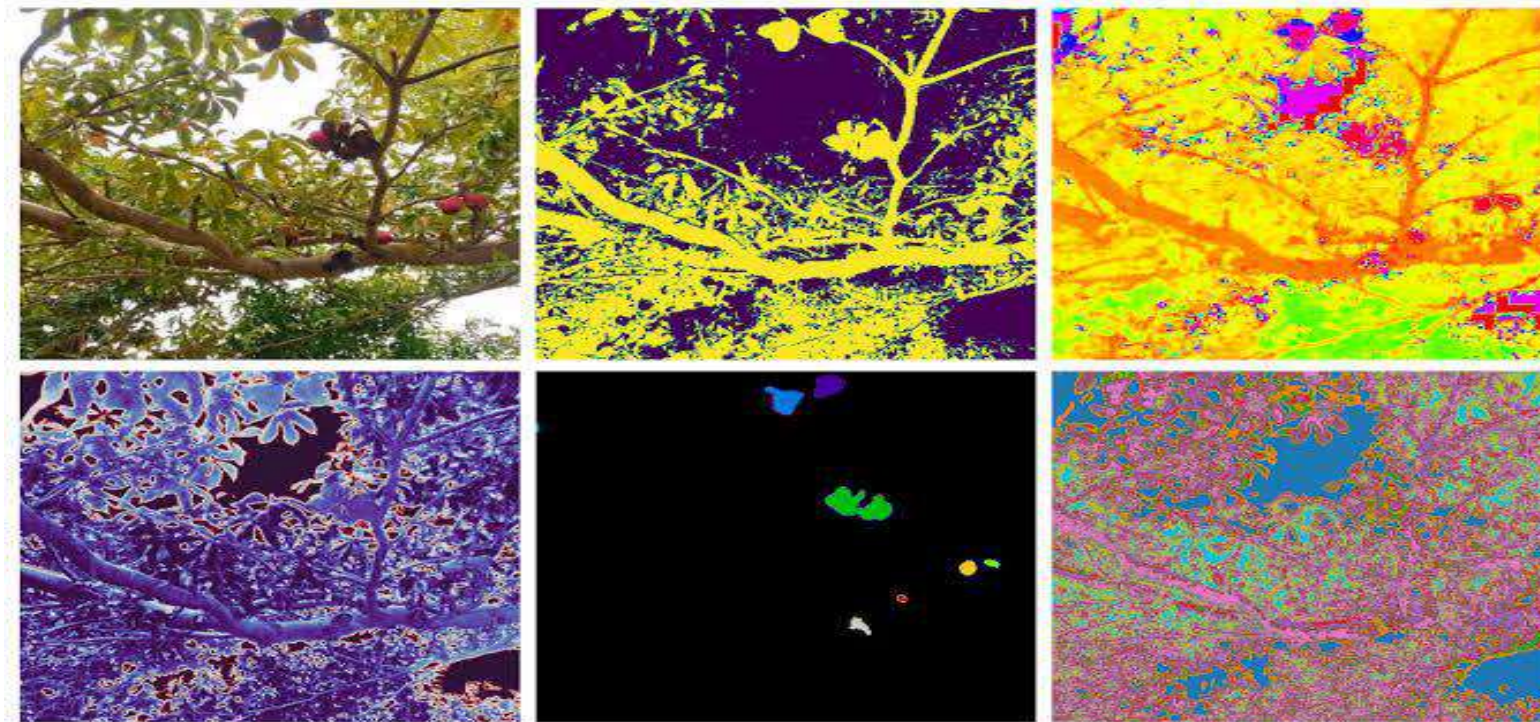
THE YOLO MODEL

- YOLO calculates the probabilities and bounding boxes for objects in a single forward pass. The way this is done is splitting the image into grids and detecting objects for each of the boxes.
- However, this is not done sequentially, and is actually implemented as a convolution operation.
- For example: if
- Input image size is (448,448,3)
- Number of classes to predict = 25 and Grid size = (7,7)
- Then, each label vector will be of length 30 (Pc, Bx, By, Bw, Bh, +25 classes).



BLOB IMAGE

- Input Video will be converted into frames and the number of frames per second depends on the processor.
- YOLO architecture needs blob images and hence normal images are converted into blob images for processing.
- When compared to the phase 1 now we are able to get better efficiency because we used YOLO V3 algorithm.



Software implementation :

Python

- Python is a widely used general-purpose, high-level programming language. Its syntax allows the programmers to express concepts in fewer lines of code when compared with other languages like C, C++ or java.
- To get started installing Python, visit Python.org and downloaded the version 3.8.3. windows x86-64 executable installer by setting correct path. Then install the libraries pandas, requests, pillow, and OpenCV in command prompt by giving the command pip install.

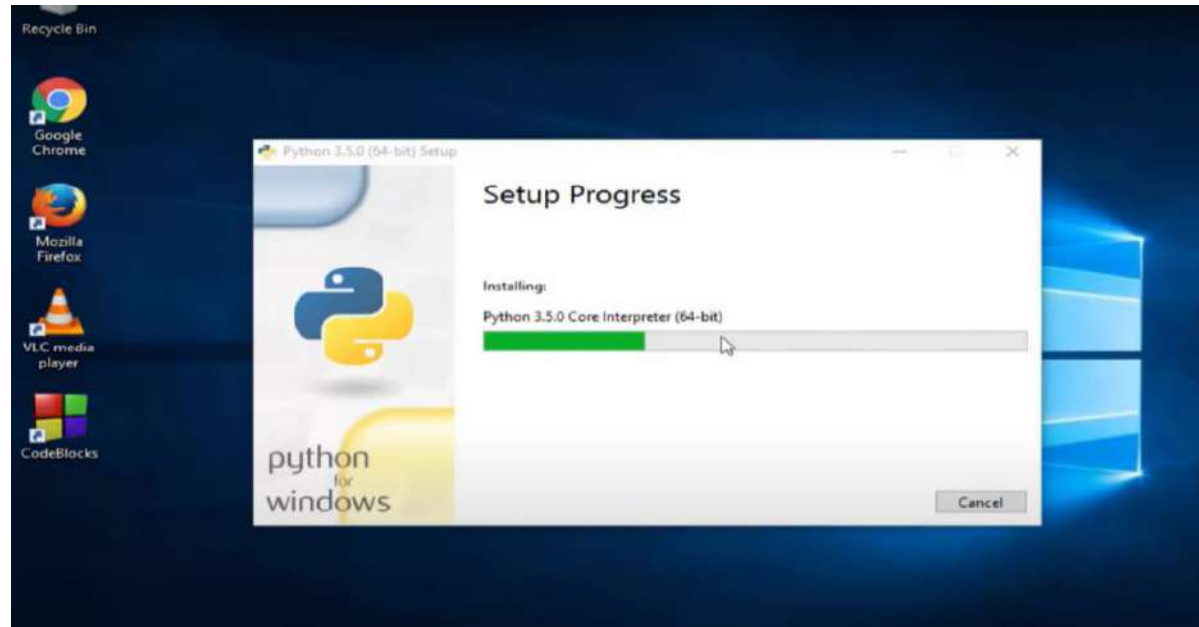


Fig: Python Installing

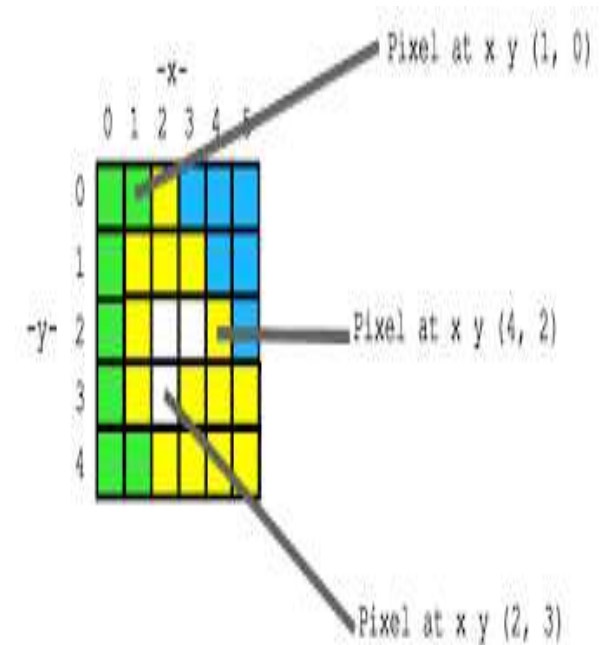
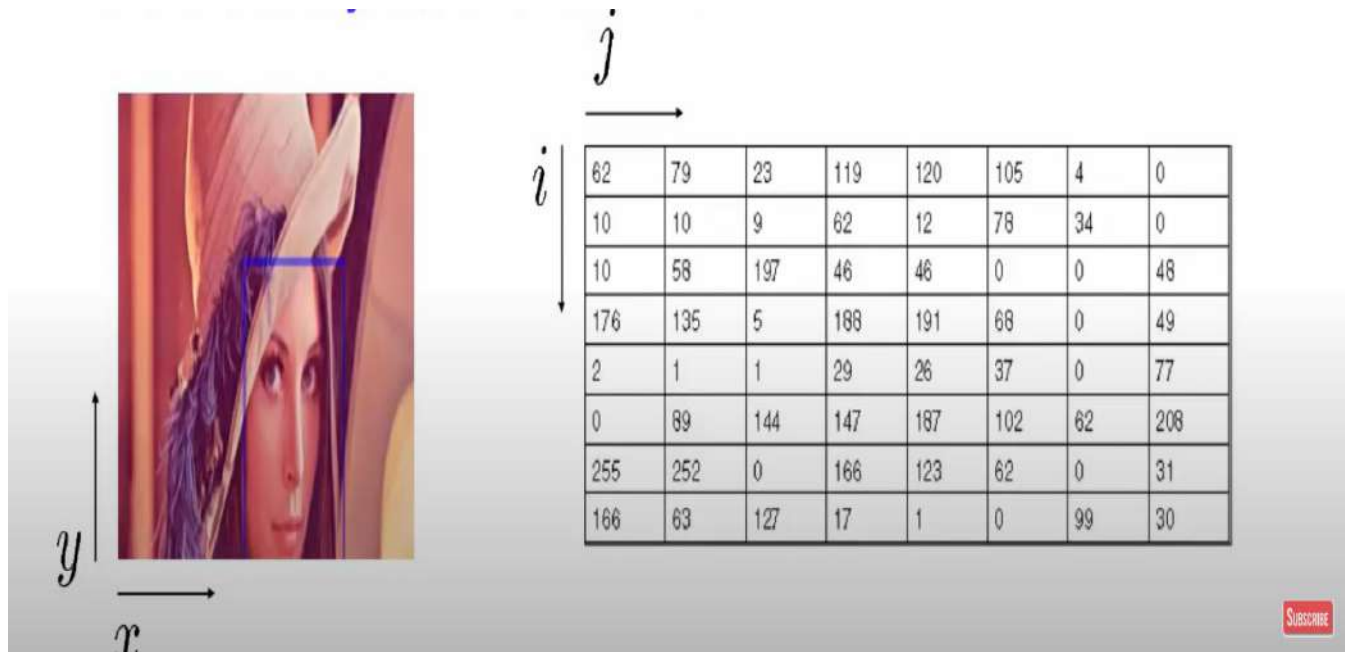
Open CV :

- Open CV is an image processing library created by Intel and later supported by Willow Garage and now maintained by Itseez.
- Available on Mac, Windows, Linux.
- Works in C, C++ and Python.
- Open Source and free.
- Easy to use and install.

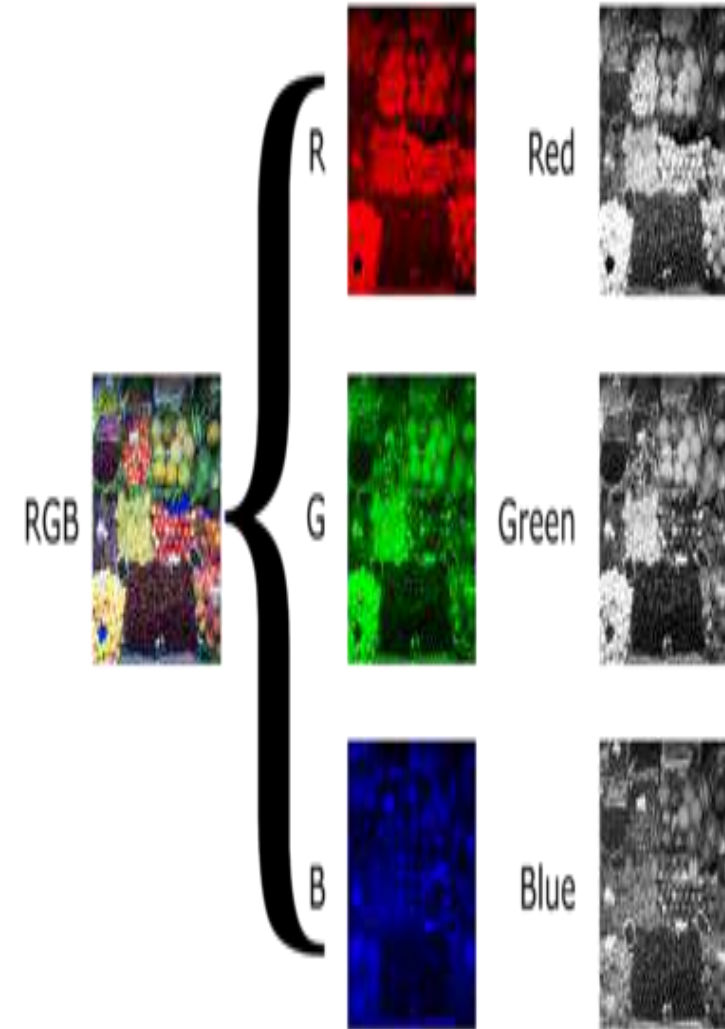


Digital Images

- Digital images are typically stored in matrix.
- There are many different file formats.



How do computers see images?



There are two types digital images

1. **Grayscale Images**
2. **Colored Images**

1.Grayscale Images:

In this each pixel represents the intensity of only one shade that is how bright or dark the pixel is. In other word it is said that it has only one channel.

2. Colored Images:

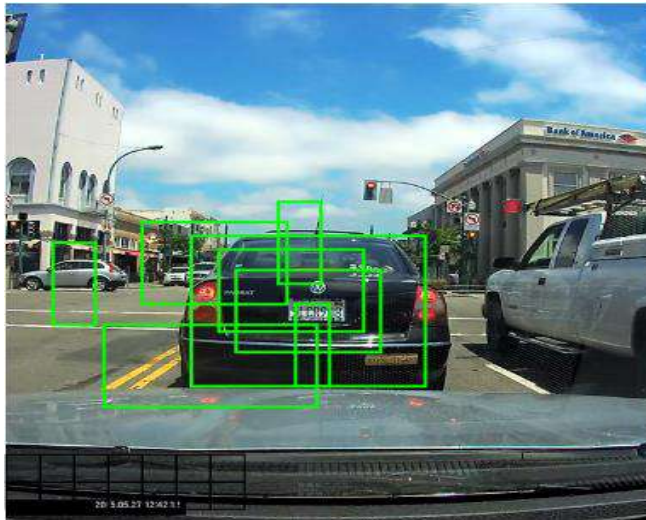
In this we have three channels they are R, G, B (Red, Green, Blue)

Numpy :

- Numpy is a highly optimized library for numerical operations.
- Array Structure is important because digital images are 2D arrays of PIXELS.
- All the OpenCV array structures are converted to numpy arrays.
- We can use more convenient indexing system rather than using for loops.

Non-max suppression and threshold filtering

- Non-max suppression makes use of a concept called “intersection over union” or IoU.
- The first step, quite naturally, is to get rid of all the boxes which have a low probability of an object being detected.
- It takes as input two boxes, and the name implies, calculates the ratio of the intersection and union of the two boxes.



Non max
supression



OCR

- Optical character recognition or optical character reader (OCR) is the electronic or mechanical conversion of images of typed, handwritten or printed text into machine-encoded text, whether from a scanned document, a photo of a document, a scene-photo or from subtitle text super imposed on an image.

COCO (Common Objects in Context)

This program uses the COCO (Common Objects in Context) class list, which has 25 object categories.

When the detected images matches the objects list in coco class, the object number gets incremented by 1 in the program.

Graphical user interface:

- A GUI (graphical user interface) is a system of interactive visual components for computer software.
- A GUI displays objects that convey information, and represent actions that can be taken by the user.

Tkinter

- It is the standard GUI library for Python. Python when combined with Tkinter provides a fast and easy way to create GUI applications.
- We create buttons in the GUI page for each applications.

Source code :

- **Install Python IDE of version 3.8.3 on Windows 10 +**
- First, import the all the messages from tkinter

```
from tkinter import *
```

```
from tkinter import messagebox
```

```
top = Tk()
```

Declare a variables and initialize the color, height, width and size

```
C = Canvas(top, bg="blue", height=250, width=300)
```

```
filename = PhotoImage(file = "resized_test.png")
```

```
background_label = Label(top, image=filename)
```

```
background_label.place(x=0, y=0, relwidth=1, relheight=1)
```

Import window from tkinter and assign the title, geometry and configuration.

```
from tkinter import *
```

```
window=tk.Tk()
```

```
root>window
```

```
window.title("project1")
```

```
window.geometry("3500x900")
```

```
window.configure(background="#E4287C")
```

Labelling will be done by declaring a variable called lbl and labelling the name with color, size, width, height of rectangle for tripleriding.

```
lbl=Label(window, text="Triple Ride and Helmet Detection" ,fg="black" ,width=25 ,height=1,font=('times', 30, 'italic bold underline'))
```

```
lbl.place(x=400,y=10)
```

Then import os and define time()

import os

def time():

os.startfile("yolo_detection_webcam1.py")

def time1():

os.startfile("Helmet_detection_YOLOV3.py")

def time2():

os.startfile("yolo_detection_images.py")

def time3():

os.startfile("yolo_detection_webcam.py") #start python file

def time4():

os.startfile("helmet.py")

def time5():

os.startfile("live1.py")

def time6():

os.startfile("yolo_detection_images4.py")

def time8():

Import the Python libraries, and declare threshold filtering and NMS (Non Maxima Suppression) values

```
import numpy as np
```

```
import cv2                                     #importing libraries
```

```
import cv2 as cv
```

```
confidenceThreshold = 0.0
```

```
NMSThreshold = 0.6                           #ranging of pixel
```

Take the dataset inputs from coco and other dataset files

```
modelConfiguration = 'cfg/yolov3.cfg'
```

```
modelWeights = 'yolov3.weights'
```

```
labelsPath = 'coco.names'
```

```
labels = open(labelsPath).read().strip().split('\n')
```

Then VideoCapture() will help to capture the input video

```
video_capture = cv2.VideoCapture("om1.mp4")
```

```
(W, H) = (None, None)
```

```
count = 0
```

writing the while loop and if loop for repeating the process of capturing video

while True:

```
    ret, frame = video_capture.read()
```

```
    fram1=frame
```

```
    #frame = cv2.flip(frame, 1)
```

if W is None or H is None:

```
    (H,W) = frame.shape[:2]
```

Non Maxima Suppression will be done in this part of the code.

```
#Apply Non Maxima Suppression
```

```
detectionNMS = cv2.dnn.NMSBoxes(boxes, confidences, confidenceThreshold, NMSThreshold)
```

```
if(len(detectionNMS) > 0):
```

```
    for i in detectionNMS.flatten():
```

```
        (x, y) = (boxes[i][0], boxes[i][1])
```

```
(w, h) = (boxes[i][2], boxes[i][3])
```

```
        color = [int(c) for c in COLORS[classIDs[i]]]
```

```
print(labels[classIDs[i]])
```

```
if labels[classIDs[i]]=="motorbike": #since my detector only has 1 class
```

```
    cv2.imwrite("tripleride//framet%d.jpg" % count, frame[y-200:y+h, x-20:x+w])
```

Finally when video capture is over, release the video capture and destroyAllWindows

```
video_capture.release()
```

```
cv2.destroyAllWindows()
```

Further declarations and assigning values are required for the video and image input and output frames.

except:

```
print("completed video")
```

def time9():

```
os.startfile("yolo_detection_images6.py")
```



```
def time7():  
    import requests  
    import base64  
    import json  
    from glob import glob  
    import pandas as pd  
    import time  
    import os  
    def ocr(IMAGE_PATH):  
        SECRET_KEY = 'sk_fa7d3dcec0363bdfb6ac3e06'  
        with open(IMAGE_PATH, 'rb') as image_file:  
            img_base64 = base64.b64encode(image_file.read())
```

```
l=set(l)

print(l)

for text in l:

    raw_data = {'date':[time.asctime( time.localtime(time.time()))],":[text]}

    #raw_data = [time.asctime( time.localtime(time.time()))],[text]

df = pd.DataFrame(raw_data)

    df.to_csv('data.csv',mode='a')

os.startfile('data.csv')

btn1=Button(window, text="Helmet input Video", command=time ,fg="blue" ,bg="orange" ,width=30 ,height=3
,activebackground = "white" ,font=('times', 18, ' bold '))

btn1.place(x=0,y=150)
```

```
btn1=Button(window, text="Detect Persons", command=time2 ,fg="blue" ,bg="orange" ,width=30 ,height=3 ,activebackground  
= "white" ,font=('times', 18, ' bold '))
```

```
btn1.place(x=500,y=150)
```

```
btn1=Button(window, text="Detect Helmet", command=time1 ,fg="blue" ,bg="orange" ,width=30 ,height=3 ,activebackground  
= "white" ,font=('times', 18, ' bold '))
```

```
btn1.place(x=980,y=150)
```

```
btn1=Button(window, text="TripleRide input Video", command=time8 ,fg="blue" ,bg="White" ,width=30 ,height=3  
,activebackground = "white" ,font=('times', 18, ' bold '))
```

```
btn1.place(x=150,y=280)
```

```
btn1=Button(window, text="Detect Persons", command=time9 ,fg="blue" ,bg="White" ,width=30 ,height=3 ,activebackground  
= "white" ,font=('times', 18, ' bold '))
```

```
btn1.place(x=900,y=280)
```

```
btn1=Button(window, text="Detect NumberPlate", command=time6 ,fg="blue" ,bg="White" ,width=30 ,height=30 ,activebackground = "white" ,font=('times', 18, ' bold '))
```

```
btn1.place(x=150,y=450)
```

```
btn1=Button(window, text="Detect Text NumberPlate", command=time7 ,fg="blue" ,bg="White" ,width=30 ,height=30 ,activebackground = "white" ,font=('times', 18, ' bold '))
```

```
btn1.place(x=900,y=450)
```

```
btn1=Button(window, text="Detect Tripleride video", command=time3 ,fg="blue" ,bg="#00ff00" ,width=30 ,height=30 ,activebackground = "white" ,font=('times', 18, ' bold '))
```

```
btn1.place(x=0,y=580)
```

```
btn1=Button(window, text="Detect helmet video", command=time4 ,fg="blue" ,bg="#00ff00" ,width=30 ,height=3  
,activebackground = "white" ,font=('times', 18, ' bold '))
```

```
btn1.place(x=500,y=580)
```

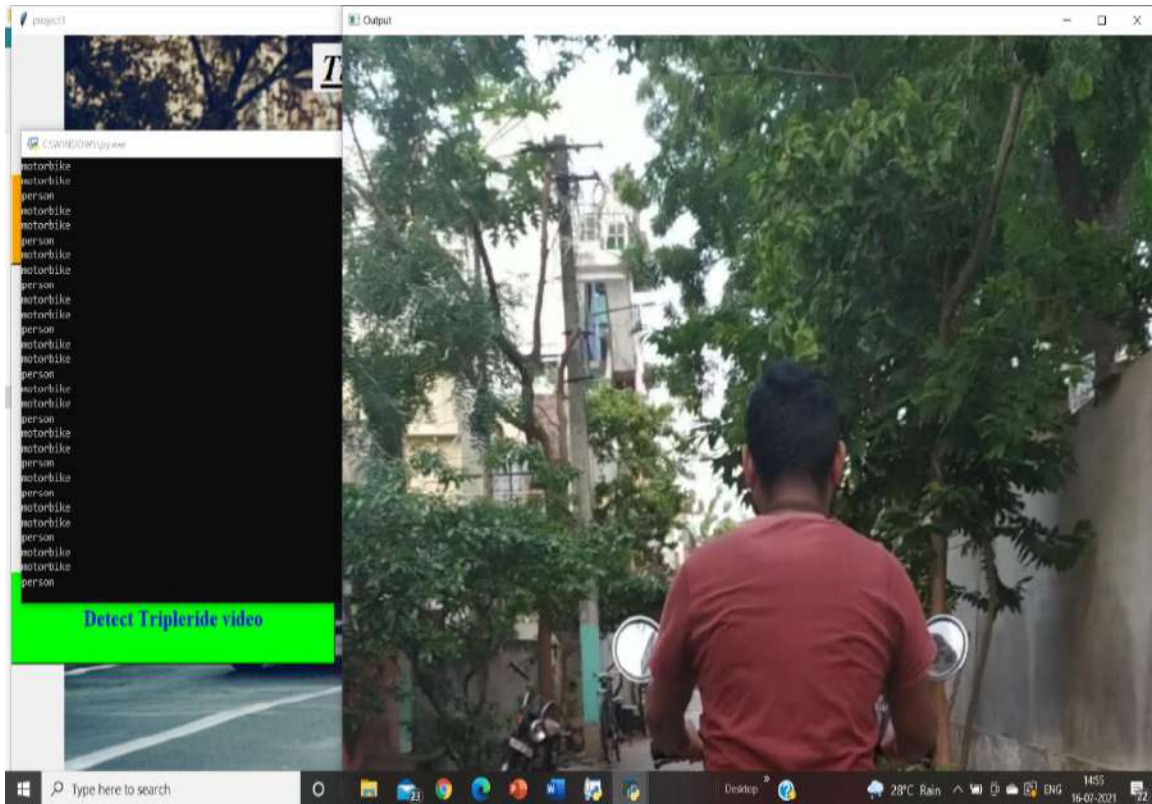
```
btn1=Button(window, text="traffic Detection", command=time5 ,fg="blue" ,bg="#00ff00" ,width=30 ,height=3  
,activebackground = "white" ,font=('times', 18, ' bold '))
```

```
btn1.place(x=980,y=580)
```

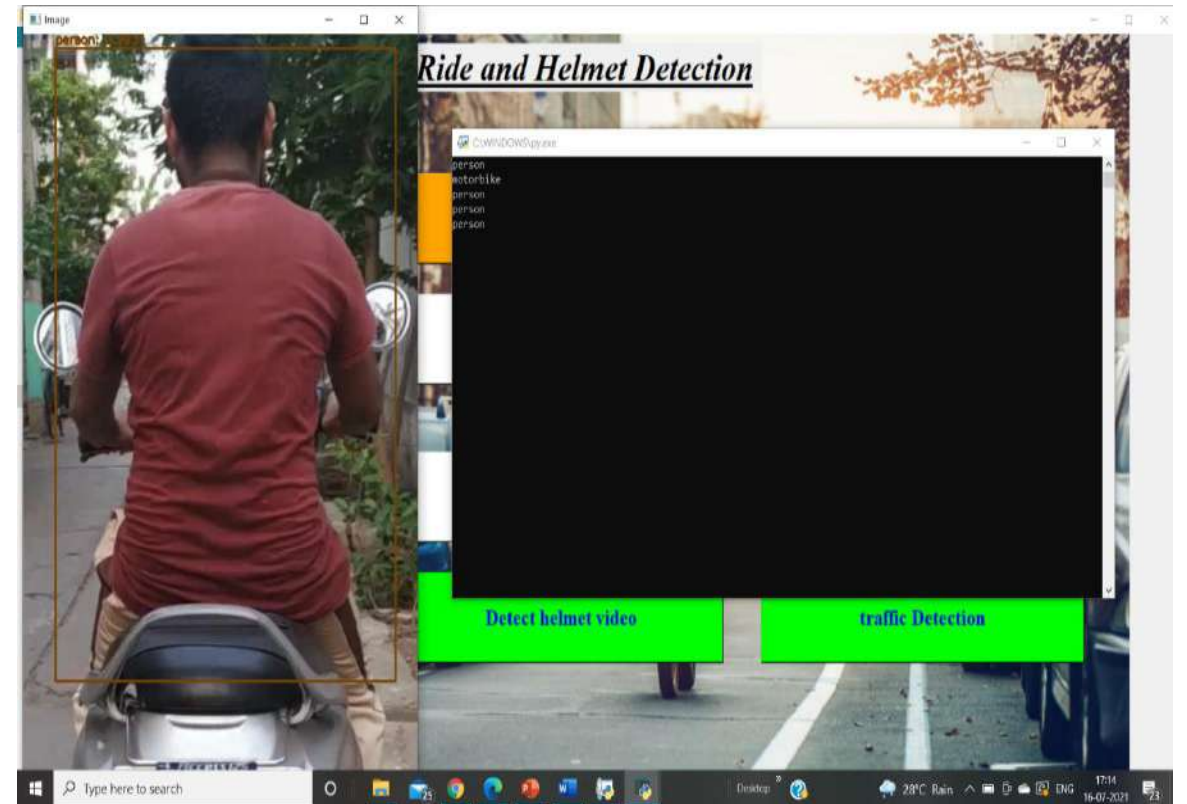
```
window.mainloop()
```

Result :

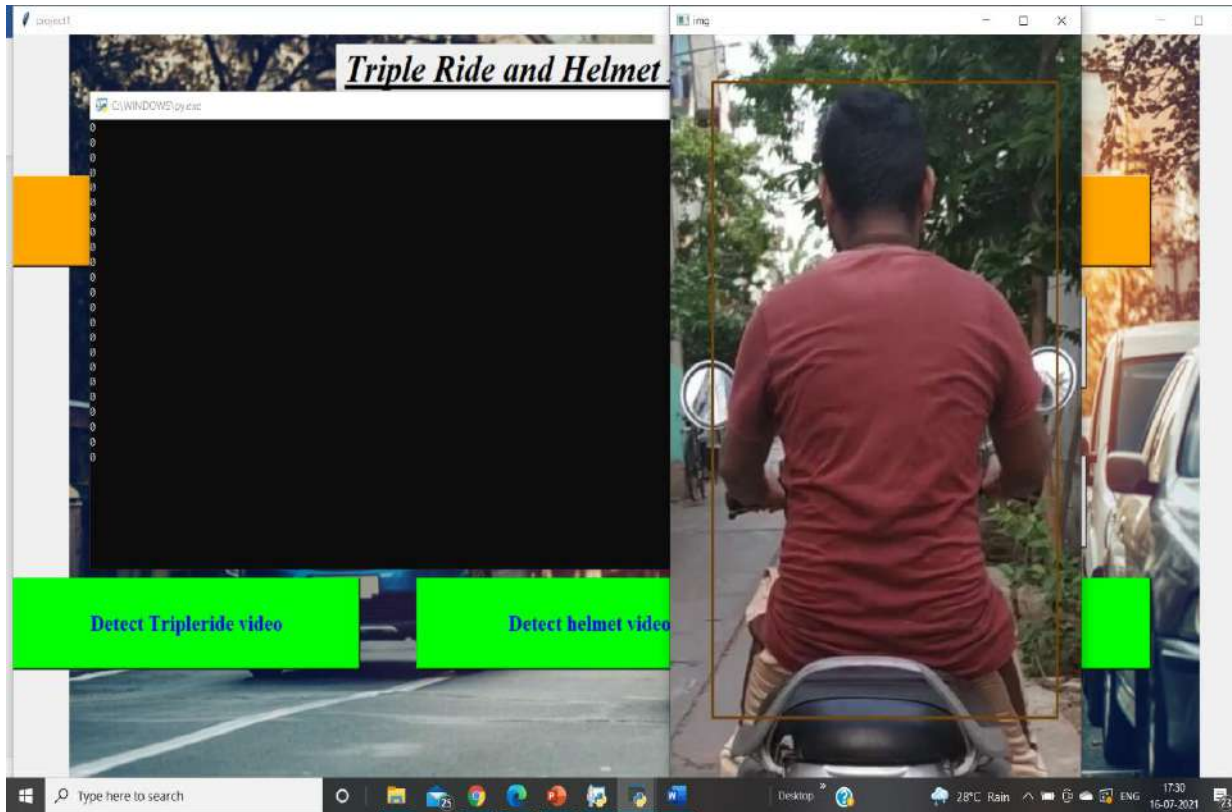
Helmet Input Video



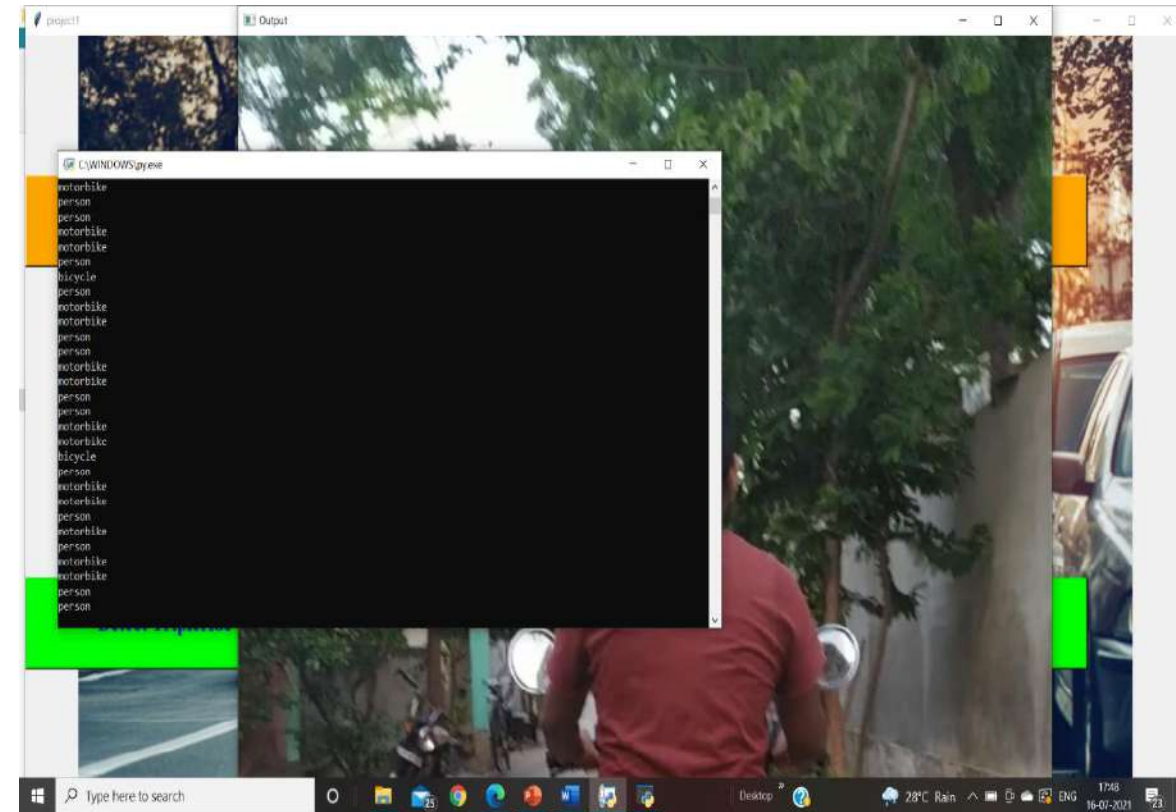
Detect persons



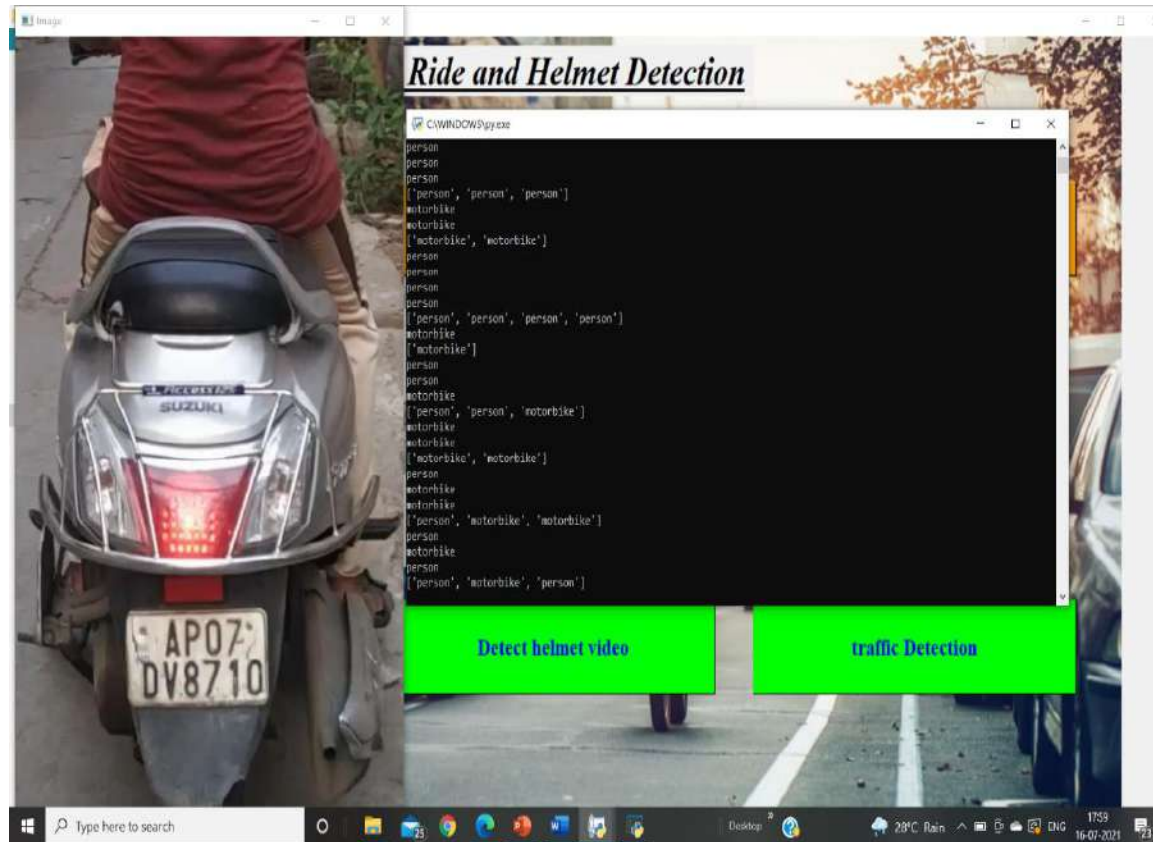
Detect Helmet



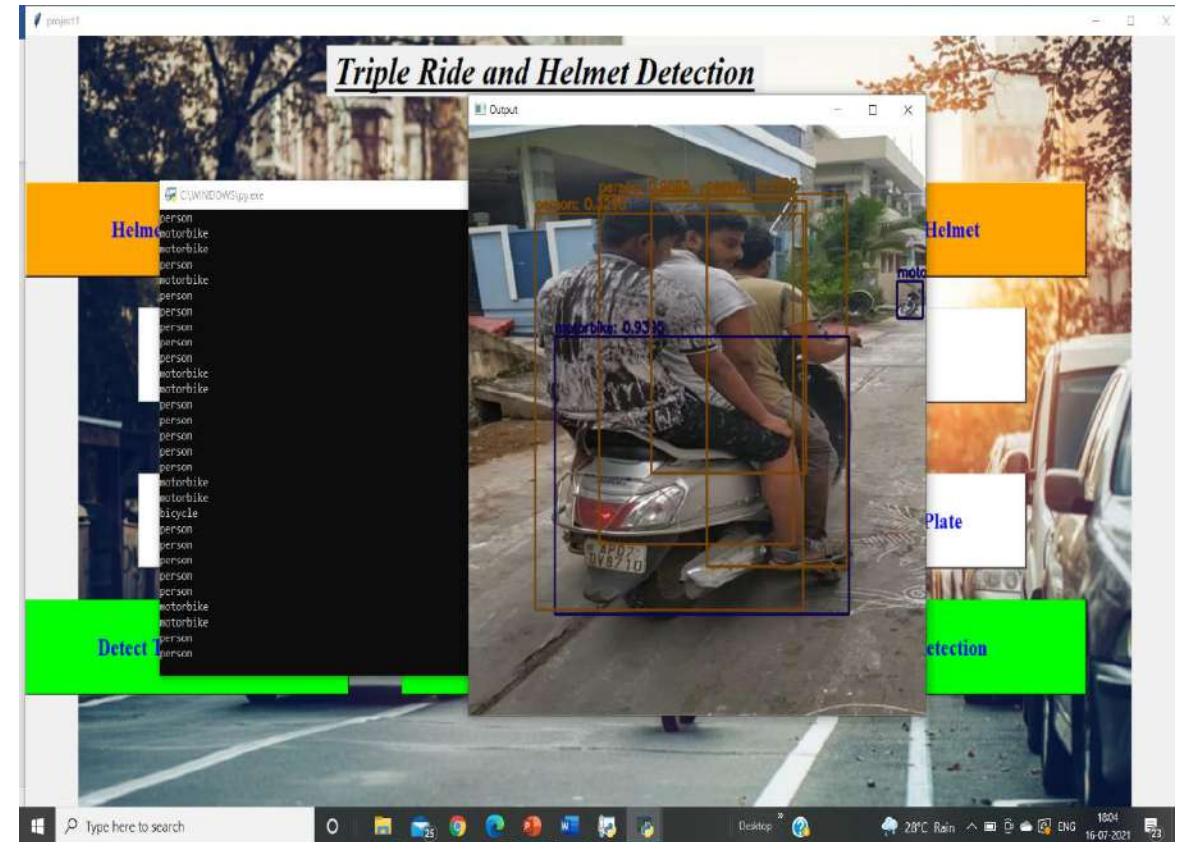
TripleRide input Video



Detect Persons



Detect Tripleride video

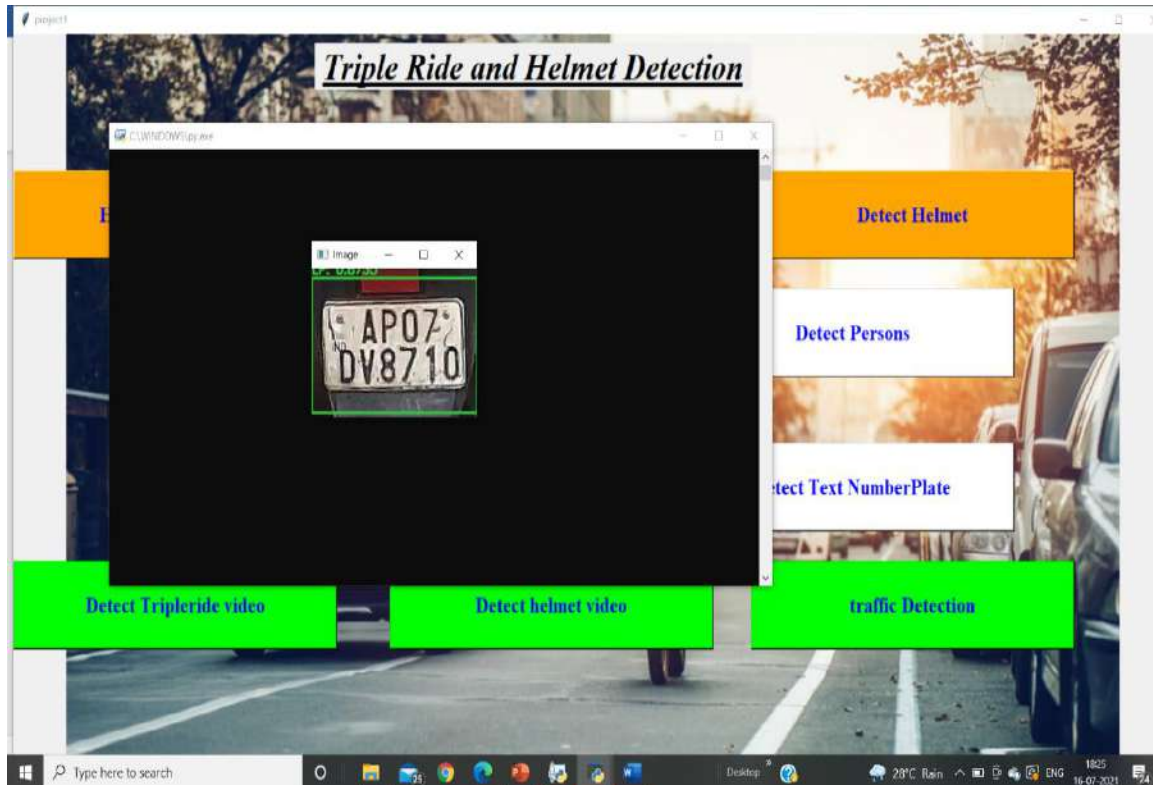


The screenshot displays a video analysis application interface. The main video frame shows a motorcycle accident scene. Several detection results are highlighted with colored boxes and text labels:

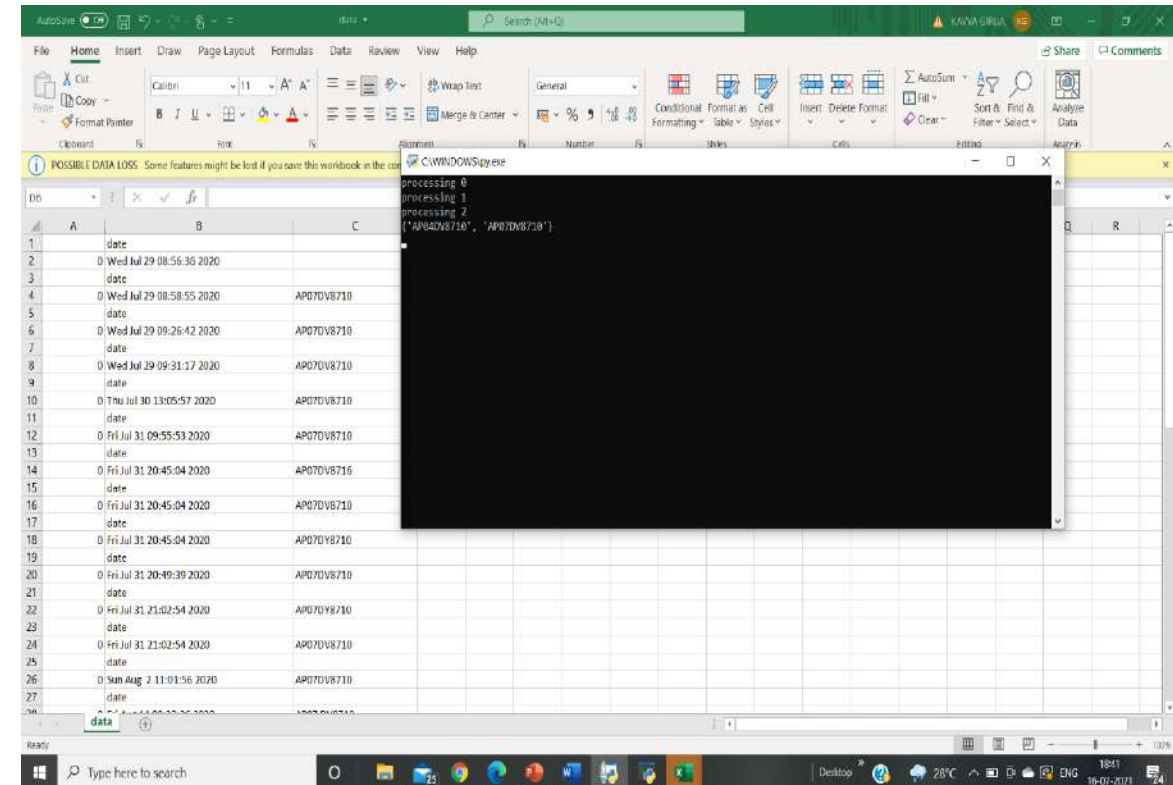
- Green box (bottom left):** Detect Triplride video
- Green box (bottom right):** traffic Detection
- Yellow box (top right):** Detect Helmet
- White box (right):** Detect Persons
- White box (right):** Text NumberPlate

A small 'Output' window is visible in the center. The background of the application is a blurred image of a street scene with a motorcycle and people.

Detect NumberPlate



Detect Text NumberPlate



Conclusion

- As when compared to other algorithm YOLO is found be more advantageous and has higher efficiency and accuracy. Hence, we use the same approach to identify triple riding.
- Since we are implementing our project using YOLO algorithm which is a CNN based approach, results are obtained at fastest speed .

Future scope

- In future we can implement the traffic rules violation system in a high profile quality. So that we can improve the traffic rules violation system.
- In future we can use camera and other sensor identify the vehicle and identify the traffic rules violation and can automatically send the images to the administrator.
- When the fine is applied the people can easily understand and they will come to know the fine for their violation.

Advantages

- Automation of this procedure is exceptionally attractive for vigorous observing of these infringements and additionally it likewise altogether lessens the measure of human resource required.
- It does continuous monitoring of traffic rules 24x7.
- No need of human intervention.
- An automated system, if employed to detect and penalize motorcycle riders without a helmet, will naturally motivate the two-wheeler riders to wear helmets.
- Reduce road accidents.

Applications

- Can be used in all traffic signals.
- Can be used in any type of streets.

References :

- 1] Automatic Motorcyclist Helmet Rule Violation Detection using Tensorflow & Keras in OpenCV - IEEE Conference Publication :
<https://ieeexplore.ieee.org/document/9087168>
- [2] <https://www.sciencedirect.com/science/article/pii/S026322411830900X>, 2019
- [3] Pattern Recognition and Machine learning by Christopher Bishop in 2006
- [4] Digital image processing by William K Pratt

THANK YOU