



# A Practical Weapon Detection System For Enhanced Security Using Artificial Intelligence

CYSF 2025 - ALVEENA ASHIQ - Grade 8 - Westmount Charter School



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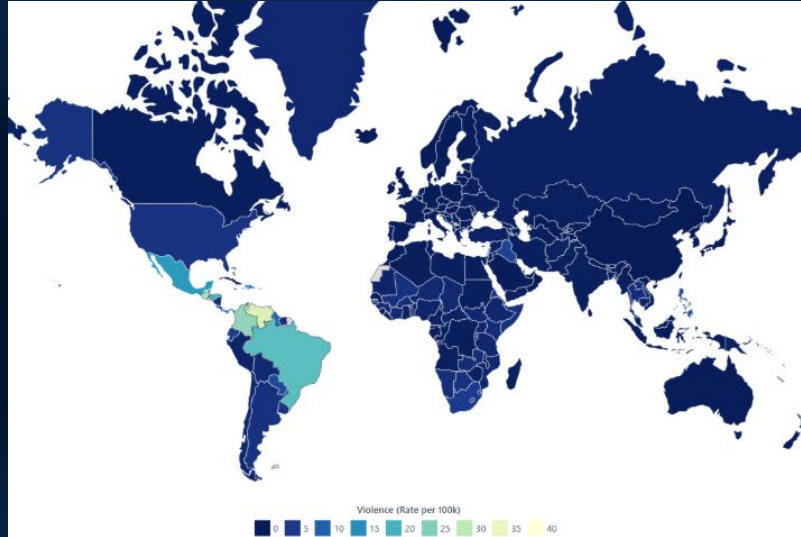


01

Problem



# Gun and Weapon Violence



Over 250,000 people died from firearms in 2019, with the majority of deaths being homicides, especially in countries like Brazil and the United States.

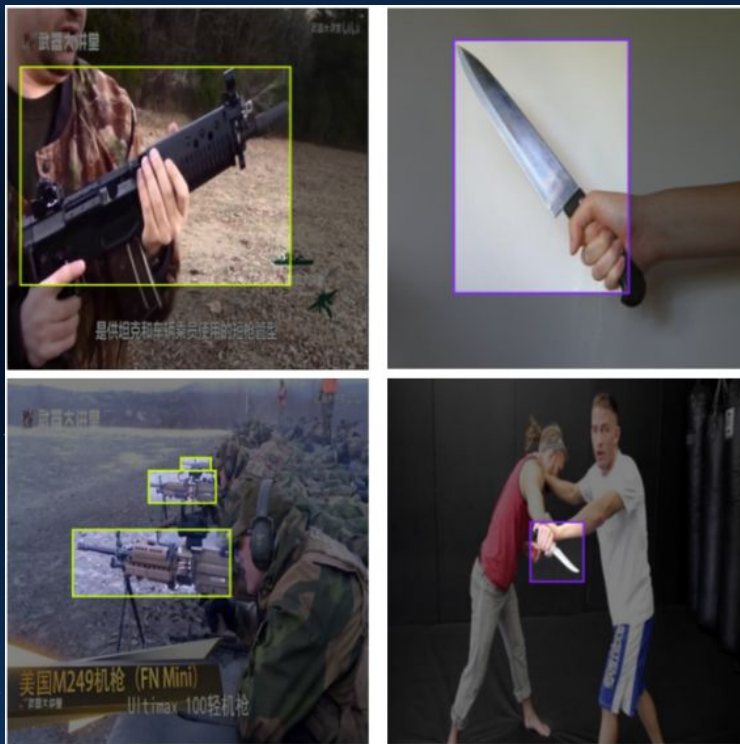


02

Method



# Object Detection



Some labelled data of guns and knives

# Development

## A Windows-based workstation

- Demonstrates the suitability of the system for a central monitoring station
- Can be connected to multiple IP security cameras to detect weapons.

## Python

- Windows, macOS, Linux, Raspbian, and more.
- NumPy, OpenCV, PyTorch, and others.
- Ideal for rapidly validating a proof of concept or building a prototype.

## Raspberry Pi-based camera system

- Equipped with weapon detection analytics
- Cost-effective solution
- does not require a central monitoring station for operation.





# YOLOv8 Models

## Nano Model

- For developing integrated weapon detection system
- Works well with minimal computational resources of Raspberry Pi
- Provides faster inference but lower accuracy.

## Large Model

- Higher accuracy at the cost of slower inference speed and increased computational demands.
- For the Windows-based workstation application
  - Typically have sufficient computational power to handle a Large model

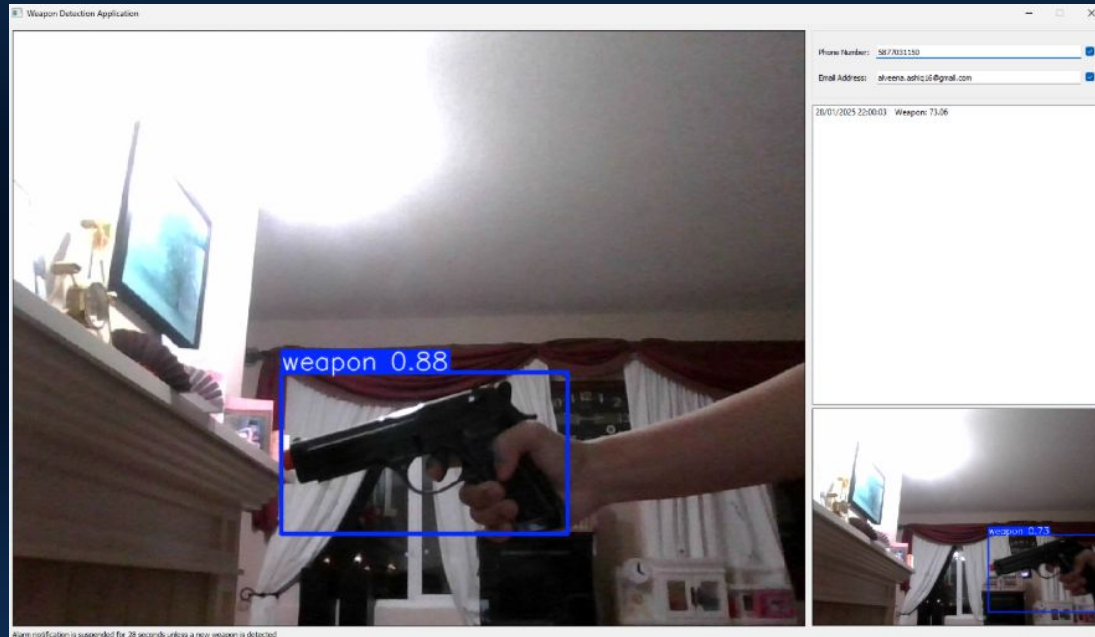




```
27
28 class VideoThread(QThread):
29     change_pixmap_signal = pyqtSignal(QVariant, arguments=['data'])
30
31     def __init__(self, channel):
32         super().__init__()
33         self.channel = channel
34         self.model = YOLO('../YOLOv8l_WeaponModel.pt')
35
36     def run(self):
37         # capture from web cam1')
38         vidcap = cv2.VideoCapture(self.channel)
39
40         success, image_BGR = vidcap.read()
41
42         while (success):
43             now = datetime.now()
44             success, image_BGR = vidcap.read()
45
46             results = self.model.predict(image_BGR, conf = 0.7)
47
48             # Visualize the results on the frame
49             annotated_frame = results[0].plot()
50
51             aDict = {"image": annotated_frame, "result": results[0], "datetime": now}
52
53             self.change_pixmap_signal.emit(aDict)
54
55         # shut down capture system
56         vidcap.release()
57
```

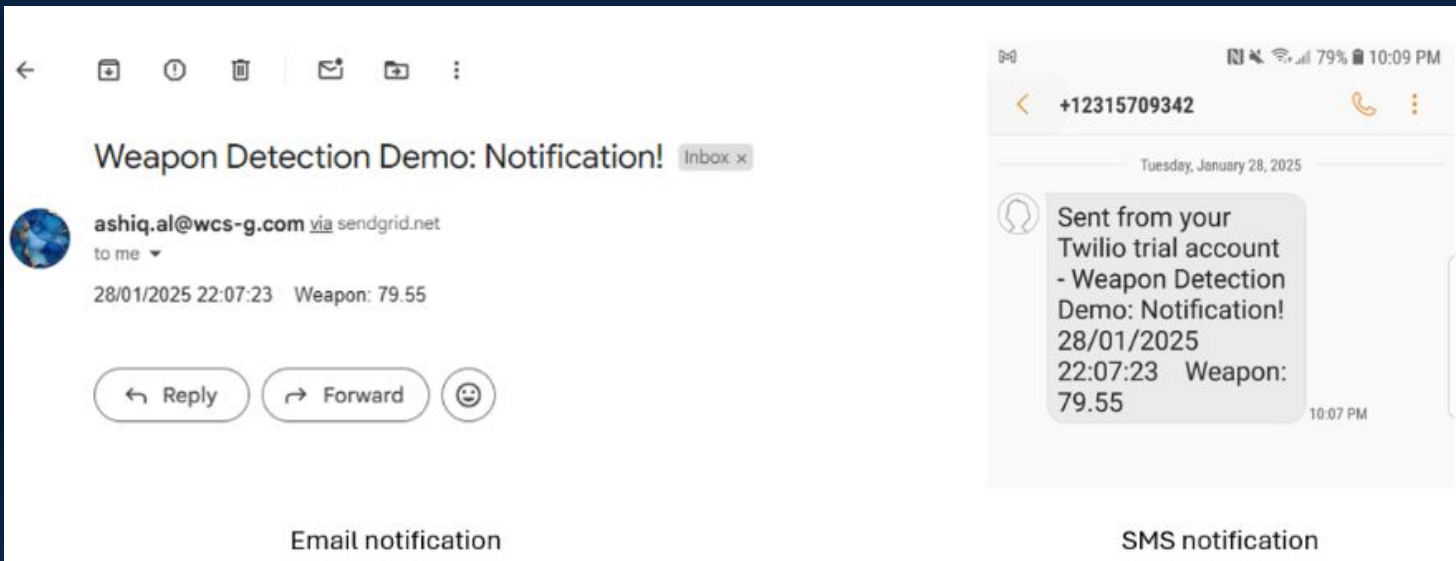
A code snippet from the developed application

# Application



Weapon detection application interface

# Notification



Real-time weapon detection notification via email and SMS

# Raspberry PI Camera Prototype



Raspberry PI camera prototype with integrated weapon detection analytics

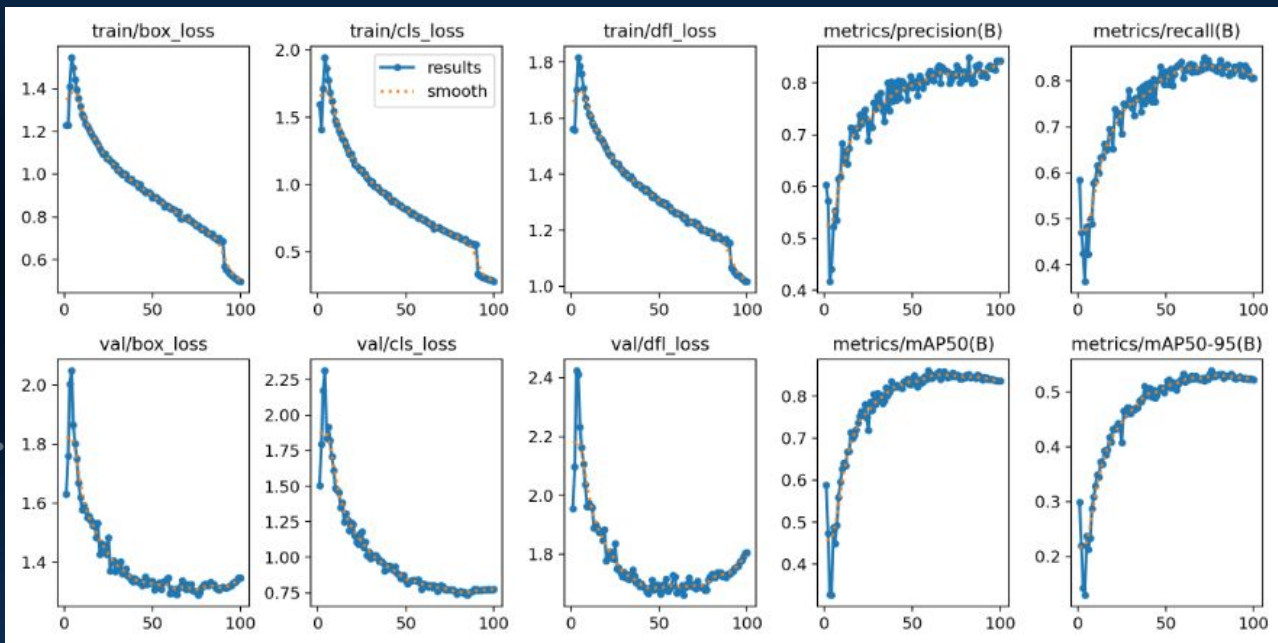




Analysis



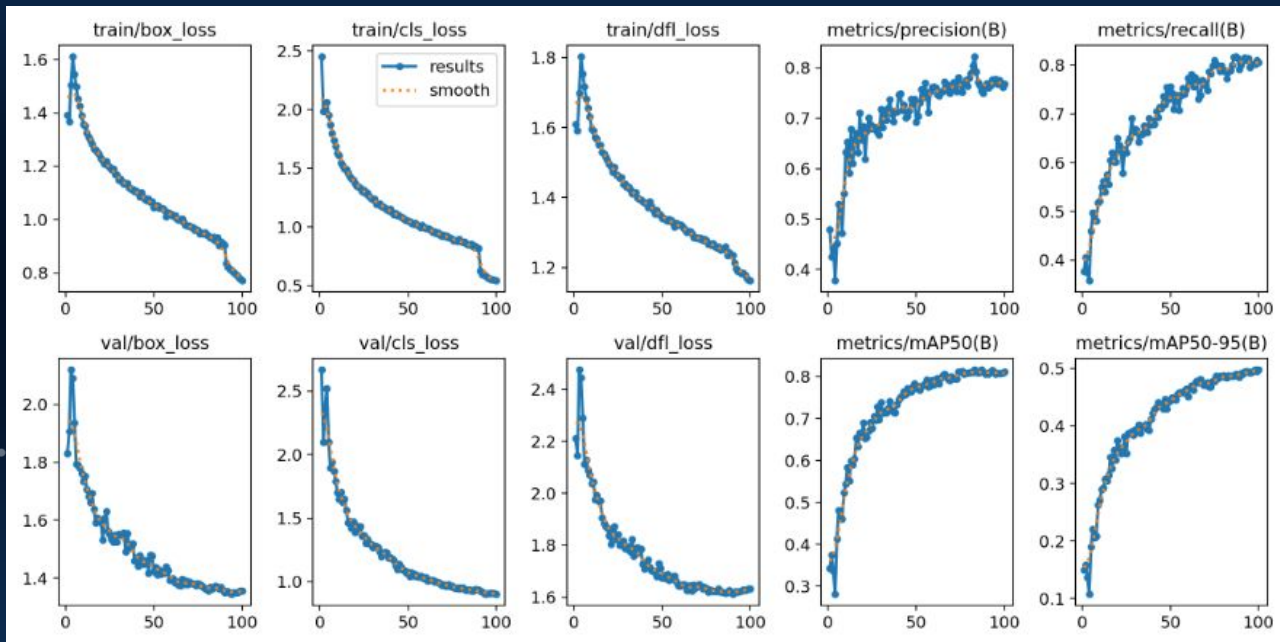
# Performance Analysis



YOLOv8 Large model training results from over 100 epochs using the weapon dataset



# Performance Analysis



YOLOv8 Nano model training results from over 100 epochs using the weapon dataset

# Performance Analysis

Model	Precision (%)	Recall (%)
YOLOv8l	82.6	82.8
YOLOv8n	76.8	80.4

Precision and recall values of the trained YOLOv8l and YOLOv8n models





04

# Conclusion



# Two weapon detection systems

A central monitoring station-based application capable of analyzing streaming video from commercial IP cameras

A cost-effective, standalone edge camera device with built-in weapon detection analytics.



# Thank you!

For listening to my CYSF presentation

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