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SCHOOL OF COMPUTER SCIENCE AND ENGINEERING

PROJECT REPORT ON MACHINE LEARNING

MINI PROJECT

Robot Fencing Project

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> Under The Guidance of Mrs Neha Tarannum

Abstract

The human detection robot is a state-of-the-art device designed to detect the presence of humans in a given area. Equipped with advanced sensors and algorithms, the robot is able to accurately detect human presence and provide real-time alerts to the user. One of the key features of the human detection robot is its ability to operate in a variety of environments, including both indoor and outdoor settings. The robot's robust design allows it to withstand harsh weather conditions and operate reliably in a wide range of temperatures. Additionally, the human detection robot is easy to use and requires minimal training to operate. The user can easily configure the robot's settings and adjust its sensitivity to accurately detect human presence in the desired area. Overall, the human detection robot is a valuable tool for any company looking to enhance security and improve safety in their premises. The project's domain is 'Robot Fencing' where the robot detects any human occurance and stops working if there is any human detection near it. This project will be based on object detection algorithm which will detect if there is any human presence around the robot .

1 Acknowledgement

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Vaibhav Kumar Aman Raj Anand Doddamani Yajas Menon

Contents

1	Acknowledgement	3
2	Chapter 1: Introduction	5
	2.1 Motivation.	6
	Literature Survey	6
	2.3 Problem Statement	10
	2.4 Applications	11
	2.5 Objectives	12
3	Chapter 2: Requirement Analysis	13
	Functional Requirement	13
	3.2 Non Functional Requirements	14
	3.3 Hardware Requirements	15
	3.4 Software Requirements	15
	Chantar 2 System Design	
4	Chapter 3 System Design	17
	4.1 Architectural Design	17
	4.2 Proposed Methodology	18
5	Chapter 4: Implementation	20
	5.1 Video Capturing	20
	5.2 Training	21
	5.3 Testing	23
6	Chapter 6: Results	24
7	Chapter 7: Conclusion	25

2 Chapter 1: Introduction

Introducing the human detection robot - a state-of-the-art device designed to accurately detect the presence of humans in a given area. Equipped with advanced sensors and algorithms, the robot is able to quickly and reliably identify human presence, providing real-time alerts to the user.

The human detection robot is an invaluable tool for any company looking to enhance security and improve safety in their premises. With its ability to operate in a variety of environments, including both indoor and outdoor settings, the robot is versatile and durable, making it ideal for use in a wide range of settings.

In addition to its versatility and durability, the human detection robot is also easy to use and requires minimal training to operate. The user can easily configure the robot's settings and adjust its sensitivity to accurately detect human presence in the desired area. This user-friendly design makes the robot accessible to individuals with a wide range of technical expertise, allowing companies to deploy the robot quickly and efficiently.

The human detection robot is not only versatile and easy to use, but it also provides real-time alerts to the user. When the robot detects human presence, it sends an alert to the user, allowing them to quickly respond to any potential security or safety concerns. This real-time notification system allows companies to react quickly to potential threats and ensures that their premises are always protected.

In conclusion, the human detection robot is a valuable tool for any company looking to enhance security and improve safety in their premises. With its advanced sensors and algorithms, versatile design, and user-friendly interface, the robot is capable of accurately detecting human presence and providing real-time alerts to the user, making it an essential component of any security system.

2.1 Motivation

There are several key motivations for using a human detection robot in a company. First and foremost, the robot can enhance security and improve safety in the company's premises. By accurately detecting the presence of humans in a given area, the robot can alert the user to potential security threats and help prevent accidents or other safety incidents.

Another key motivation for using a human detection robot is its versatility and durability. The robot is able to operate in a variety of environments, including both indoor and outdoor settings, and is built to withstand harsh weather conditions. This makes it an ideal tool for use in a wide range of settings, including industrial facilities, warehouses, and other environments where human presence may be important to monitor.

Additionally, the human detection robot is easy to use and requires minimal training to operate.

This user-friendly design makes the robot accessible to individuals with a wide range of technical expertise, allowing companies to deploy the robot quickly and efficiently.

Overall, the human detection robot is a valuable tool for any company looking to enhance security and improve safety in their premises. Its advanced sensors and algorithms, versatile design, and user-friendly interface make it an essential component of any security system.

2.2 Literature Survey

The paper is titled "Human Detection Robot using YOLOv5 Algorithm". According to the abstract of the paper, the main focus of the research was to design and develop a human detection robot using the YOLOv5 object detection algorithm. The paper describes the design and implementation of the robot, including the hardware and software components used. It also discusses the performance of the robot in detecting humans, and compares the results to other object detection algorithms. The paper concludes that the YOLOv5 algorithm is a suitable choice for implementing a human detection robot, as it has a high detection accuracy and a fast processing speed. It also suggests that further work could be done to improve the performance of the robot, such as by adding additional sensors or using a larger training dataset. Overall, this research paper provides a detailed overview of the design and implementation of a human detection robot using the YOLOv5 object detection algorithm. It may be of interest to researchers and practitioners working in the fields of robotics, computer vision, and artificial intelligence.[1]

The paper titled "A comparative study of YOLOv5 models' performance for image localization and classification." According to the abstract of the paper, the main focus of the research was to compare the performance of different YOLOv5 models for image localization and classification tasks. The authors trained and evaluated several YOLOv5 models on the COCO dataset, a widely used dataset for object detection tasks. They compared the models' performance in terms of the mean average precision (mAP) metric, which measures the accuracy of object detection.

The paper found that the YOLOv5-SPP model had the best performance, followed by the YOLOv5-Tiny and YOLOv5-Large models. The authors also observed that the YOLOv5 models had a faster processing speed compared to other state-of-the-art object detection models such as RetinaNet and Faster R-CNN.

Overall, this research paper provides a detailed comparison of the performance of different YOLOv5 models for image localization and classification tasks. It may be of interest to researchers and practitioners working in the fields of computer vision and artificial intelligence.[2]

In this paper titled "Finding RTSP addresses for IP cameras, NVRs, and DVRs." This paper Provides information and instructions on how to locate the RTSP (Real-Time Streaming Protocol) address of a network-based camera, such as an IP camera, NVR (Network Video Recorder), or DVR (Digital Video Recorder). The RTSP address is a URL that allows a client device, such as a computer or smartphone, to connect to the camera and view the video stream in real-time. The article explains that the RTSP address may be listed in the camera's documentation or on the manufacturer's website, and provides steps for accessing the address through the camera's web interface.

The article also provides tips for troubleshooting common issues that may arise when trying to access the RTSP address, such as incorrect login credentials or network connectivity problems. It also offers suggestions for using third-party software or online services to view the video stream, if the RTSP address is not available or not working.[3]

In this paper Titled "Configure IP Camera on Network." . This paper provides information and instructions on how to set up and configure an IP (Internet Protocol) camera on a network.

An IP camera is a type of network-based camera that is connected to a network using an Ethernet cable or WiFi connection. The article explains the steps involved in setting up an IP camera, including connecting the camera to the network, configuring the camera's settings, and accessing the video stream.

The paper also provides information on how to access the camera's web interface, which is a web-based control panel that allows the user to configure the camera's settings and view the video stream. It describes the various settings and options available in the web interface, such as resolution, frame rate, and network settings.[4]

In this research paper titled "YOLOv7: Trainable bag-of-freebies sets new state-of-the-art for real-time object detectors." According to the abstract of the paper, the main focus of the research was to evaluate the performance of the YOLOv7 object detection algorithm on various benchmarks. YOLOv7 is a real-time object detection algorithm that has achieved state-of-the-art performance on several benchmarks.

The authors of the paper trained and evaluated the YOLOv7 algorithm on the COCO, VOC, and Open Images datasets, which are widely used for object detection tasks. They found that YOLOv7 outperformed other state-of-the-art object detection algorithms in terms of the mean average precision (mAP) metric, which measures the accuracy of object detection.

The paper also describes the "trainable bag-of-freebies" approach used in YOLOv7, which is a set of techniques that can be applied to improve the performance of the algorithm. The authors found that the use of these techniques significantly improved the performance of YOLOv7 on the COCO dataset.[5]

In this Paper Titled "Application of an Improved YOLOv5 Algorithm in Real-Time Detection of Foreign Objects by Ground Penetrating Radar." According to the abstract of the paper, the main focus of the research was to evaluate the performance of the YOLOv5 object detection algorithm for detecting foreign objects using ground penetrating radar (GPR) data. GPR is a non-invasive method of detecting objects and features underground using radar waves.

The authors of the paper used an improved version of the YOLOv5 algorithm, which they developed by adding additional layers to the model and fine-tuning it on a dataset of GPR images. They evaluated the performance of the improved YOLOv5 algorithm on a dataset of GPR images and found that it had a high accuracy in detecting foreign objects.

The paper also discusses the potential applications of the improved YOLOv5 algorithm in real-time detection of foreign objects using GPR data, such as in the fields of archaeology, civil engineering, and environmental monitoring. [6]

To build a model which detect	ts Humans from the imag	es captured using an in	dustry provided
IP camera and printing the val	lues in 0's and 1's.		

2.4 Applications

There are several key applications for a human detection robot in a company. One of the primary uses of the robot is to enhance security and improve safety in the company's premises. By accurately detecting the presence of humans in a given area, the robot can alert the user to potential security threats and help prevent accidents or other safety incidents.

One specific application of the human detection robot is in industrial facilities and warehouses. In these environments, the robot can be used to monitor areas where human presence may be important, such as restricted access areas or areas with hazardous materials. By accurately detecting human presence in these areas, the robot can help prevent unauthorized access and ensure that employees are not exposed to potential dangers.

In addition to its security and safety applications, the human detection robot can also be used for research and development purposes. By providing accurate data on human presence in a given area, the robot can help researchers study human behavior and develop new technologies and strategies for improving safety and security.

Another applications of the human detection robot is a versatile and valuable tool for any company looking to enhance security and improve safety in their premises. Its advanced sensors and algorithms, versatile design, and user-friendly interface make it an essential component of any security system.

2.5 Objectives

The objectives for a human robot detection system for a company may vary depending on the specific needs and goals of the company. Some possible objectives for such a system could include:

- \bullet To build a model that can detect the Region of interest (ROI) from the given input video.
- \bullet To build a model that can detect Humans from the detected region of interest (ROI)
- To store the detected Humans in the local database, in excel format there by reducing manual intervention.

3 Chapter 2: Requirement Analysis

Requirement analysis provides an overview of the entire system with purpose, scope, definitions, acronyms, abbreviations, references, and overview of the software requirements specification.

3.1 Functional Requirement

This section gives specific information about the functional requirements of the Robot Fencing Project. Functional requirements define a system or its component along with the main functionality of the system that defines the system as a whole unit. Below are the functional requirements of the system.

- Network connectivity: The IP camera must be connected to the network and accessible from the device running YOLOv5. This might require setting up the IP camera with a static IP address or configuring your network to allow access to the camera.
- Video streaming: The IP camera must be able to stream video to the device running YOLOv5. This might require setting up the camera to use a specific video streaming protocol, such as RTSP or MJPEG.
- Video input: YOLOv5 must be able to receive video input from the IP camera. This
 might require using a library or framework that is capable of accessing the camera's
 video stream and feeding it into YOLOv5.
- Output display: You may want to display the output of YOLOv5 on a screen or monitor. This might require using a library or framework that is capable of displaying images or video with bounding boxes and labels overlaid on them.
- Performance: Depending on the resolution and frame rate of the video stream, you
 may need to optimize the performance of YOLOv5 to ensure that it can process
 the video in real-time. This might involve adjusting the model's hyperparameters or
 using a more powerful device to run the model.
- Security: You may need to consider security measures when using an IP camera, such
 as setting up a secure connection to the camera and protecting the device running
 YOLOv5 from unauthorized access.

3.2 Non Functional Requirements

This section gives specific information about the Non-functional requirements of the Robot Fencing Project. Non-functional requirements deal with the performance attributes of the system. It justifies how the system should exist in terms of its usability, performance, reliability, maintainability, etc. Below are the Non-functional requirements of the system.

- Accuracy: The model should be able to detect objects with a high degree of accuracy, minimizing the number of false positives and false negatives.
- Speed: The model should be able to process the video stream in real-time, without introducing significant delays or lag.
- Robustness: The model should be able to handle a wide range of input conditions, such as different lighting conditions and object poses, without degrading performance.
- Scalability: The model should be able to handle a large number of object classes and detections without significantly slowing down.
- Compatibility: The model should be compatible with the IP camera and the device running YOLOv5, as well as any libraries or frameworks used to access the camera's video stream and display the model's output.
- Maintainability: The model should be easy to maintain and update, with clear documentation and a straightforward configuration process.
- Security: The model should be secure, with measures in place to prevent unauthorized access and protect sensitive data.

3.3 Hardware Requirements

This section gives information about the Hardware Requirements for the Robot Fencing Project. This is related to the devices on which the algorithm is compatible. Just to run the algorithm, minimal hardware requirements are employed. Below are the Hardware requirements of the system.

- 8GB Ram.
- Intel i5 or i7 processor.
- 256 GB SSD and if there is a graphics card then it would be better

3.4 Software Requirements

This section gives information about the Software Requirements for the Robot Fencing Project. This is related to the operating systems on which this Project is compatible and technologies used.

3.4.1 Operating System to run

Machine Learning algorithm requires a platform or specifically, operating system to execute the program. This section includes various operating systems that are compatible. Below are the operating systems that share the software compatibility with the current project.

- Windows.
- Mac OS.
- Linux.

3.4.2 Technologies Used

This section includes various technologies that proved helpful to build the system effectively. Algorithms indeed need some powerful technologies to stand out in terms of efficiency, maintainability and every other aspect. The different technologies used in building the system are as below.

- Google Colab.
- Jupyter Notebook.

4 Chapter 3 System Design

In this Module This will give us the basic View of the System Design and Proposed Methodoloy(i.e Models Used in this Project).

4.1 Architectural Design

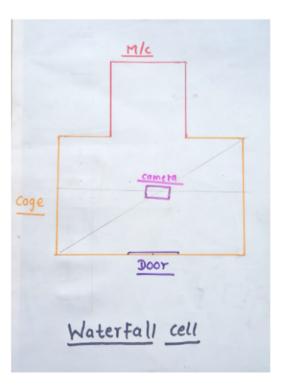


Figure 1: A Basic Process of the system

Above is the Architectural Design of Project . There will be a cage in which a robot is Working . The Robot Doesn't know to stop when there is an Human presence .

So There will be an IP camera attached at the Top of the cage . So Our Model will take the video as an input provided by the IP camera and will tell the robot whether there is an Human presence or not .

If there is a Human presence inside the cage the robot will stop working and will again start moving when there is no Human inside the cage.

4.2 Proposed Methodology

In this Section we will discuss about the models which we have tried for this Project.

4.2.1 Proposed Methodology:1 Yolo v5 Model

YOLO (You Only Look Once) is a popular real-time object detection system that is fast and accurate.YOLO v5 is the latest version of YOLO, which was released in 2021. It is an improvement over the previous versions, and it is faster and more accurate than its predecessors.

The YOLO v5 model works by dividing the input image into a grid of cells, and each cell is responsible for predicting a set of bounding boxes and class probabilities. The model makes a prediction for each cell in the grid, and these predictions are combined to form the final object detections

YOLO v5 is trained on large datasets of annotated images, and it uses a variant of the Darknet architecture, which is a convolutional neural network (CNN). The model is trained to predict the class probabilities and bounding boxes of objects in an image, and it is optimized using an objective function that measures the accuracy of the predictions.

There are many applications for YOLO v5, including object tracking, video surveillance, and autonomous vehicles. It is also used in a variety of industries, including retail, security, and transportation.

Model	size (pixels)	mAP ^{val} 50-95	mAP ^{val} 50	Speed CPU b1 (ms)	Speed V100 b1 (ms)	Speed V100 b32 (ms)	params (M)	FLOPs @640 (B)
YOLOv5n	640	28.0	45.7	45	6.3	0.6	1.9	4.5
YOLOv5s	640	37.4	56.8	98	6.4	0.9	7.2	16.5
YOLOv5m	640	45.4	64.1	224	8.2	1.7	21.2	49.0
YOLOv5I	640	49.0	67.3	430	10.1	2.7	46.5	109.1
YOLOv5x	640	50.7	68.9	766	12.1	4.8	86.7	205.7
YOLOv5n6	1280	36.0	54.4	153	8.1	2.1	3.2	4.6
YOLOv5s6	1280	44.8	63.7	385	8.2	3.6	12.6	16.8
YOLOv5m6	1280	51.3	69.3	887	11.1	6.8	35.7	50.0
YOLOv5l6	1280	53.7	71.3	1784	15.8	10.5	76.8	111.4
YOLOv5x6 + TTA	1280 1536	55.0 55.8	72.7 72.7	3136 -	26.2 -	19.4 -	140.7 -	209.8

Figure 2: Yolo v5 Model and its version

We have Used Yolo v5 Model For our Project.

5 Chapter 4: Implementation

5.1 Video Capturing

We are Taking Video as an input from an IP Camera Provided by the industry .

```
import cv2
import numpy as np

# cap = cv2.VideoCapture('rtsp://admin:Admin@123@192.168.1.108:554/cam/realmonitor?channel=1&subtype=0')
cap = cv2.VideoCapture(0)
net = cv2.dnn.readNetFromONNX("yolov5x.onnx")
file = open("coco.txt","r")
classes = file.read().split('\n')
print(classes)
```

Figure 3: Video Capturing

This code imports the cv2 (OpenCV) and numpy modules, which are used for computer vision tasks and numerical operations, respectively. The cv2.VideoCapture function is used to open a video stream, either from a file or from a device such as a webcam (indicated by the argument 0). The video stream is stored in the cap variable.

The cv2.dnn.readNetFromONNX function is used to load a pre-trained object detection model stored in ONNX format. This model is stored in the net variable.

The file variable is used to open the coco.txt file, which contains a list of class labels for the object detection model. The read() method is used to read the entire contents of the file, and the split(' ') method is used to split the contents into a list of strings, with each string being a class label. The resulting list of class labels is stored in the classes variable.

5.2 Training

```
while True:
    img = cap.read()[1]
    if img is None:
        break
    img = cv2.resize(img, (1000,600))
    blob = cv2.dnn.blobFromImage(img,scalefactor= 1/255,size=(640,640),mean=[0,0,0],swapRB= True, crop= False)
    net.setInput(blob)
    detections = net.forward()[0]

# cx,cy , w,h, confidence, 80 class_scores
# class_ids, confidences, boxes
```

Figure 4: Detection

This code reads frames from the video stream and processes them using an object detection model.

The while loop will continue to execute until the break statement is encountered. Inside the loop, the cap.read() function is used to read the next frame from the video stream. The [1] index is used to access the frame itself, which is stored in the img variable. If the img variable is None, it means that there are no more frames to read and the loop will exit.

The cv2.resize function is used to resize the frame to a specific width and height (in this case, 1000x600).

The cv2.dnn.blobFromImage function is used to preprocess the image for input to the object detection model. It resizes the image, scales the pixel values, and performs other operations to prepare the image for input to the model.

The net.setInput function is used to set the image as the input to the object detection model. The net.forward function is then used to run the model on the input image and generate detections. The detections are stored in the detections variable as a multidimensional array. The Below Image is for the video that we have sent after video capturing are of continuous human detection, which means the code will always run and will never end.

```
classes_ids = []
confidences = []
boxes = []
rows = detections.shape[0]

img_width, img_height = img.shape[1], img.shape[0]
x_scale = img_width/640
y_scale = img_height/640
```

Figure 5: Detection

5.2.1 Reducing Overfitting

Here we are checking the confidence that in the range from 0 to 1 if the value in greater then 0.5 then only it is detecting the person properly

```
for i in range(rows):
    row = detections[i]
    confidence = row[4]
    if confidence > 0.5:
        classes_score = row[5:]
    ind = np.argmax(classes_score)
    if classes_score[ind] > 0.5:
        classes_ids.append(ind)
        confidences.append(confidence)
        cx, cy, w, h = row[:4]
        x1 = int((cx- w/2)*x_scale)
        y1 = int((cy-h/2)*y_scale)
        width = int(w * x_scale)
        height = int(h * y_scale)
        box = np.array([x1,y1,width,height])
        boxes.append(box)

indices = cv2.dnn.NMSBoxes(boxes,confidences,0.5,0.5)
```

Figure 6: Checking

5.3 Testing

This is to create frames from a video, which will run and the video will continue until the 'q' key is pressed.

```
for i in indices:
    x1,y1,w,h = boxes[i]
    label = classes[classes_ids[i]]
    conf = confidences[i]
    text = label + "{:.2f}".format(conf)
    cv2.rectangle(img,(x1,y1),(x1+w,y1+h),(255,0,0),2)
    cv2.putText(img, text, (x1,y1-2),cv2.FONT_HERSHEY_COMPLEX, 0.7,(255,0,255),2)

cv2.imshow("VIDEO",img)
    k = cv2.waitKey(10)
    if k == ord('q'):
        break
```

Figure 7: Testing

6 Chapter 6: Results

Using the YOLO v5 model for human detection in a robot would likely yield good results, as the model is designed to be fast and accurate at detecting a wide variety of objects in real-time. However, the specific results might as well depend on a number of factors, including the quality of the training data used to train the model, the performance of the hardware on which the model is running, and the specific application for which the model is being used.

In general, the YOLO v5 model should be able to detect humans with high accuracy, provided that it has been trained on a sufficient amount of annotated data containing human images. It is also important to consider the specific environment in which the robot will be operating, as this can impact the model's performance.

7 Chapter 7: Conclusion

Our project was able to detect Human using the IP camera given and basede on that it was able to give results of 0's if there is no Human presence captured by the IP camera and 1's if there was a Human presence detected by the IP camera .

IP camera is connected with the laptop and it is capturing the live frame .



Figure 8: Setup

Frame that IP Camera Captured



Figure 9: Frame of IP Camera

Through IP camera we have trained the model for detecting a person . So this is the frame where the Persons Are Getting Detected.



Figure 10: Detection of person

After detecting the person our model will give 1 as a output and if no person is there then it will give 0 as output.

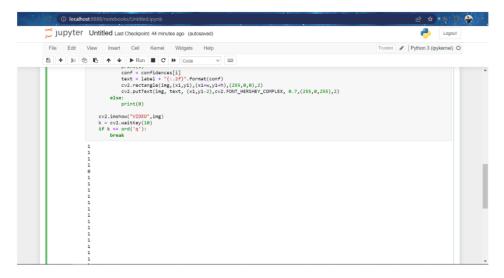


Figure 11: Final Output

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