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Blind Assistance System using Image Processing

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Abstract: Eye diseases usually cause blindness and visual impairment. According to the World Health Organization, around 40 million people are blind, while another 250 million have some form of visual impairment. They come across many troubles in their daily life, especially while navigating from one place to another on their own. They often depend on others for help to satisfy their day-to-day needs. So, it is quite a challenging task to implement a technological solution to assist them. Several technologies have been developed for the assistance of visually impaired people. One such attempt is that we would wish to make an Integrated Machine Learning System that allows the blind victims to identify and classify real-time objects generating voice feedback and distance. Which also produces warnings whether they are very close or far away from the thing.

Keywords: Blindness, Visual impairment, Machine Learning, Real-time objects

I. INTRODUCTION

The fast progress of data and organized technology has advanced from the Internet to applying innovations in life. One of the technologies to consider is objected acknowledgment innovation, later known as object detection. This term denotes a capacity to identify the shape size of diverse objects, and the device's camera catches their position. The practice of detecting real-world object instances in still photos or videos, such as a car, bike, TeleVision, flowers, and humans, is known as object detection. It lets us recognize, localize, and detect many things inside an image, giving us a better overall understanding of the scene. Image retrieval, security, surveillance, and sophisticated driver assistance systems are all examples of areas where it's applied (ADAS).

Developing accurate Machine Learning Models capable of identifying and localizing multiple objects in a single image has long been a significant challenge in computer vision. However, thanks to recent advances in Deep Learning, developing Object Detection applications is now easier than ever. TensorFlow's Object Detection API is an open-source framework built on top of TensorFlow that makes building, training, and deploying object detection models simple.

Detection of objects can be accomplished in a variety of ways. According to the World Health Organisation(WHO), over 40 million people worldwide are blind, with another 250 million who have some vision impairment. They face a lot of trouble and constant challenges in Navigation, especially when they are on their own. They need to often depend on someone to get their fundamental daily needs met. So, it is a very challenging task to make a mechanical arrangement for them which is most significant. One of our project's goals is to create an integrated Machine Learning Framework. This will allow visually impaired persons to recognize and classify daily things with voice help. Which in turn calculate distance and issue warnings if the person is too close or too far away from the object. Obstacle detection devices can be built using the same framework.

We'll concentrate on Deep Learning Object Detection in this Object Detection project because TensorFlow is based on Deep Learning. Each Object Detection Algorithm works somewhat differently, but they all follow the same basic principles.

Feature Extraction: They use their hands to extract features from input images and utilize these features to identify the image's class. MATLAB, OpenCV, Viola-Jones, and Deep Learning are just a few examples. Tensors are multidimensional arrays that extend the functionality of two-dimensional tables to data with a higher dimension. TensorFlow has numerous properties that make it suitable for Deep Learning. So, without spending any time, let's look at how we can use TensorFlow to develop Object Detection.

COCO dataset comprises around 330K annotated images for Common Objects in Context. Now you must choose a model because you must make a crucial trade-off between speed and accuracy. The main motto for object detection is to find things, drawing rectangular bounding box-like structures around them with distance. Object detection applications are emerging in numerous diverse areas of counting, recognizing people, checking crops, and real-time applications in sports.

Many methods and techniques are introduced to solve the problems of visually impaired people.

This paper gives a compelling presentation on object detection and analyzing the gesture of an object using computer vision and machine learning.

This paper proposed a well-known computer technology part of image processing and computer vision that focuses on detecting objects in computerized pictures or videos. Face detection, vehicle calculator, and character recognition are just a few of the object detection applications with high criteria.

Object detection can be used for a variety of purposes, such as recovery and surveillance. Other essential concepts used in object detection, like using the OpenCV library of python 2.7 progressing in the exactness and effectiveness of object detection, are displayed.

This paper described that everyone wants to live independently, especially the disabled ones. Over the past few decades, technology has helped disabled ones control their livelihood. In this study, an assisting system is propped for the blind using YOLO for the object detection within images and video streams based on deep neural networks to make precise detection, and OpenCV under Python using Raspberry Pi3. The result obtained indicates the proposed approach in providing blind users the capability to travel in unfamiliar indoor and outdoor environments through an object identification model and user-friendly device.

With the rise of more up-to-date and current developments, the world of innovation has prospered at a rapid rate over the last decade. Our lives have become faster due to the use of more recent advances. The rapid advancement of information and arranged innovation has progressed from the internet and mechanization frameworks, which were initially used for regulatory workplaces and mechanical and commercial applications, to the apparatus of those advances all over life. They began to consider the use of portable gadgets, apps, and versatile systems in natural checking, machine automation, smart home, etc. In order to advance computer vision frameworks, efficient and precise object recognition is essential. The introduction of machine learning and deep learning methods has dramatically increased the precision for object location. The project aims to integrate an Android application for object recognition and localization to achieve high accuracy and real-time performance.

The proposed system aims to create a visual aid image processing system for visually impaired people in which the user accepts speech commands. Its functionality addresses the identification of objects and signs. Further, the proposed system will help the visually impaired person manage day-to-day activities and navigate their surroundings.

The paper intends to incorporate cutting-edge object detection techniques to achieve high accuracy and real-time performance. In this paper, we use Python in conjunction with a TensorFlow-based approach to solving the problem of object detection from start to finish. The resulting system is quick and precise. A TensorFlow-based application for an Android mobile device is built to detect objects using the device's built-in camera, specifically:

The framework is built up in such a way that an android application (assuming you're executing it on an Android gadget) can record real-time outlines and transfer them to the program's background, where all the computations take place.

- 1) The video stream is sent and received as an input in the application's background, where it is tested and detected using accurate metrics by the COCO DATASETS object detection model.
- 2) After testing with voice modules, the path of the object will be transformed into standard voice notes, which can be delivered to blind people for help.
- 3) In addition to object discovery, we have used an alarm framework to calculate an estimate. If the Blind Person is exceptionally close to the diagram or is far away in a more secure location, it will produce voice-based results in addition to distance units.

The main objective is to identify objects and signboards to help visually impaired persons manage everyday activities. This study will assist blind people by employing speech commands to detect items using image processing techniques and provide audio output to help them navigate around barriers. This study will detect various well-known signboards, such as "Washroom," and will notify the blind person as soon as the sign is recognized.

II. REVIEW OF LITERATURE

- 1) The current methodologies for detecting models, as well as the standard datasets, were discussed in this paper. This work discussed several detectors, such as one-stage and two-stage detectors, which aided in the analysis of various object detection methods and gathered some classic and innovative applications. There were also some branches relating to object detection identified. In addition, several trends in development have been identified better to follow the set of art algorithms and subsequent processes.
- 2) This paper presented a completely convolutional network based on regions. For precise and efficient object detection, R-FCNN is used. As a result, this work can readily use ResNets as fully convolutional image classifier backbones for object detection. For object detection, this research offered a simple but effective R-FCNN architecture. When compared to the quicker R-FCNN, this approach obtains the same accuracy. As a result, it was easier to incorporate state-of-the-art picture classification backbones.
- 3) This Challenge serves as a reference point for object classification and detection. More than 100 item types and 1 million photos were categorized and detected in this work. The method for collecting enormous amounts of data is described in this publication. The most efficient algorithm for this data, as well as the successes and failures of other algorithms, were also discussed.

- 4) This study revealed that oriented gradient grids outperform the present feature set for human recognition.
- 5) According to this article, as object recognition technology has evolved, various technologies have been applied in autonomous vehicles, robots, and industrial facilities. However, the advantages of these procedures are not reaching the visually impaired, who are in severe need of them. This study proposed an object identification system for the blind based on deep learning technologies. A voice guidance technique is also used to advise visually impaired people about the position of objects. The You Only Look Once (YOLO) technique is used in the object identification deep learning model. A voice announcement is synthesized using text-to-speech (TTS) to make it easier for the blind to acquire information about items. As a result, it employs an effective object-detection system that aids the blind in locating objects within a given space.

III. METHODOLOGY

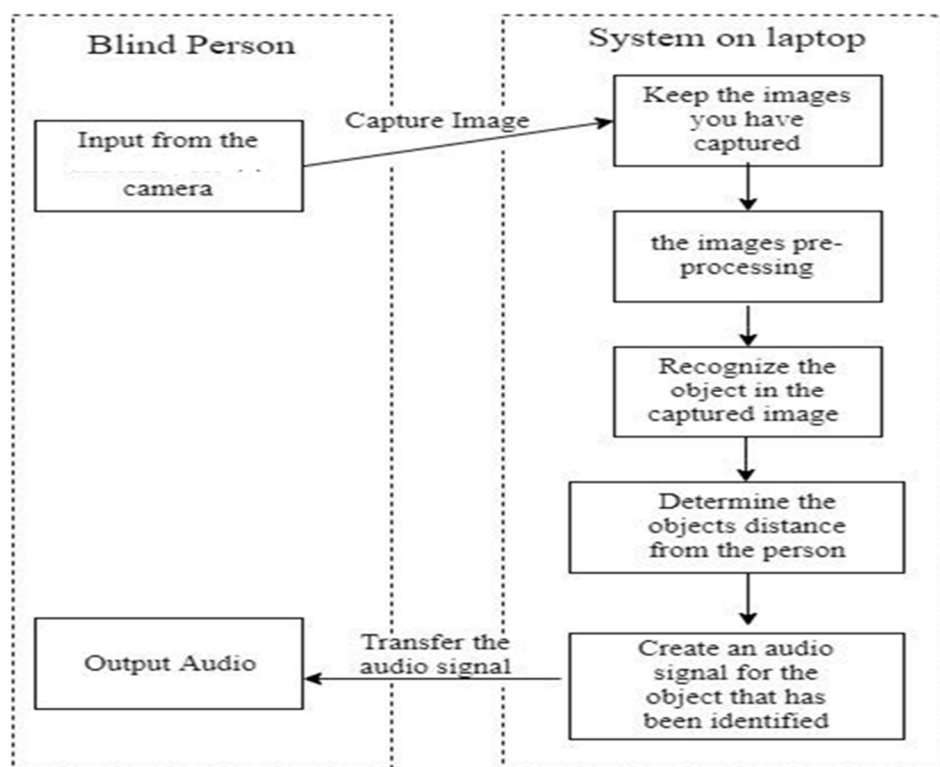


Fig 1. System Methodology

A. Overview of Technologies

TensorFlow API:



Fig 2. Logo of Tensorflow

We have implemented by using TensorFlow APIs. The advantage of using APIs is that it provides us with a set of standard operations. So, we need not write the code for a program from scratch. APIs offer us convenience, and hence they are time savers, beneficial, and efficient. The TensorFlow object detection API is generally a structure built for creating deep learning networks that solve the problem of object detection. There are so many trained models in their framework, and they refer to it as 'Model Zoo.' This 'Model Zoo' includes a collection of the COCO dataset, the KITTI dataset, and the Open Images Dataset.

TensorFlow Object Detection API depends on the libraries mentioned:

- 1) Protobuf 3.0.0
- 2) Python-tk
- 3) Pillow 1.0
- 4) lxml
- 5) tf-slim
- 6) slim
- 7) Jupyter notebook
- 8) Matplotlib
- 9) Tensorflow (1.15.0)
- 10) Cython
- 11) contextlib2
- 12) cocoapi

B. Models

Now, a bunch of pre-trained models is with Tensorflow. You can use any one of them. They are pretty good and depending upon your system specifications you can choose one. For a faster accuracy, you can go with SSD DETECTION, and for better accuracy, you can go with MASK RCNN, but most of the system shows smooth performance with SSD Mobile Net DETECTION. So, I'll elaborate on SSD ALGORITHM.

C. SSD Architecture

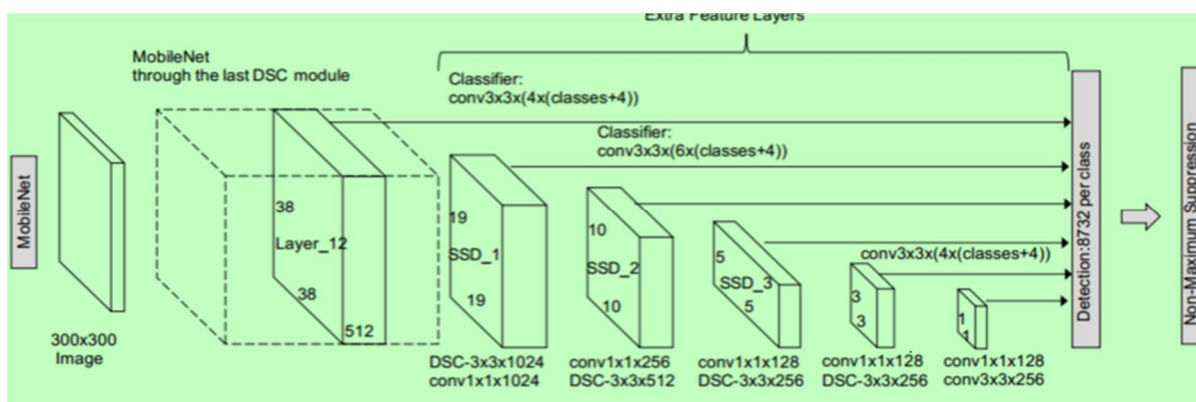


Fig 3. SSD Architecture

SSD has two components: an SSD head and a backbone model.

The backbone model is a trained image classification network as a feature extractor. Like ResNet, this is typically a network trained on ImageNet from which the final fully connected classification layer has been removed.

The SSD head is nothing but one or more convolutional layers added to the backbone. The outputs are explained as the bounding boxes and classes of objects in the spatial location of the final layer's activations. As a result, we have a deep neural network that can extract the meaning of the input image while maintaining its spatial structure at a lower resolution.

For an input image, the backbone results in 256 7x7 feature maps in ResNet34. SSD classifies the image using a grid and grid cell responsible for detecting objects in the picture region. Detecting objects means predicting the class and location of an object within that region.

D. Anchor Box

It is possible to designate multiple anchor boxes to each grid cell in SSD. These set anchor boxes are predefined, and each one is responsible for size and shape within a grid cell. The matching phase is used by SSD while training so that there's an appropriate match to anchor box with bounding boxes of each ground truth object within an image. The anchor box with the highest degree of overlap with an object is responsible for predicting that object's class and location. This attribute is used to prepare the web and anticipate the observed things and their places after the network has been trained. Each anchor box is specified with an aspect ratio and a zoom level. Well, we know that all things are not square. Some are shorter, some are very long, and some are wider by varying degrees. The SSD architecture allows predefined aspect ratios of the anchor boxes to account for this. The different aspect ratios can be specified using the ratios parameter of the anchor boxes associated with each grid cell at each zoom/scale level.

E. Zoom Level

The size of the anchor boxes does not have to be the same as the grid cell. Within a grid cell, the user may locate both more minor and more significant things. The zooms option is used to determine how much the anchor boxes should be scaled up or down for each grid cell.

F. Mobile Net

This model is based on the ideology of the mobile net model based on depthwise separable convolutions, and it forms a factorized Convolutions. This mobile net converts basic standard convolutions into depth-wise convolutions. These one \times one convolutions are also called pointwise convolutions. For MobileNets to work, these depth-wise convolutions apply a general single filter-based concept to each input channel. These pointwise convolutions use a one \times one convolution to merge with the outputs of the depthwise convolutions. As a standard convolution, both filters combine the inputs into a new set of outcomes in one single step. The depth-wise identifiable convolutions split this into two layers — a separate layer for the filtering purpose and the other separate layer for the combining purpose. This factorization methodology has the effect of drastically reducing the computation and that of the model size.

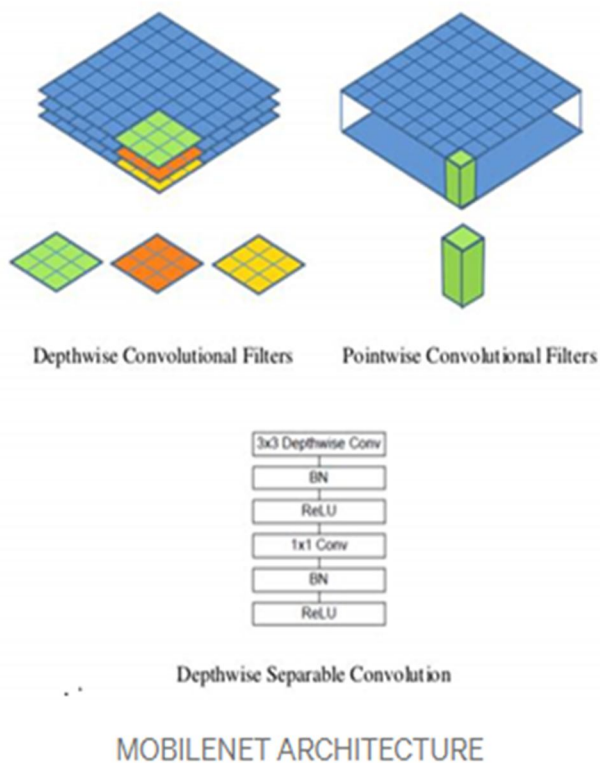


Fig 4. Mobilenet Architecture

G. Depth estimation

Depth estimation or extraction feature is nothing but the techniques and algorithms which aim to obtain a representation of the spatial building of a scene. In simpler words, it is used to calculate the distance between two objects. Our prototype is used to assist blind people, which aims to issue warnings to blind people about the hurdles coming on their way. We need to find how much distance the obstacle and person are located in any real-time situation. After the object is detected, a rectangular box is generated around that object.



Fig 5. Depth Estimation

If that object takes up the majority of the frame, the approximate distance between the thing and the human is determined based on specified limitations. The following code is used to recognize items and return space and location information.

```
(boxes, scores, classes, num_detections) = sess.run([boxes, scores, classes, num_detections], feed_dict={image_tensor: image_np_expanded})
```

Here, we have established a Tensorflow session comprised of Crucial Features for Detection. So, for further analysis, iteration is done through the boxes. Boxes are an array inside of a collection. So, for iteration, we need to define the following conditions.

```
for i,b in enumerate(boxes[0]):  
boxes[0][i][0] – y axis upper start coordinates  
boxes[0][i][1] – x axis left start coordinates  
boxes[0][i][2] – y axis down start coordinates  
boxes[0][i][3] – x axis right start coordinates
```

Index of the box in boxes array is represented by i. Analysis of the box's score is done by index. It is also used to access class. Now the width of the detected object is measured. This is done by asking the width of an object in terms of pixels.

```
apx distance = round(((1 - (boxes[0][i][3] - boxes[0][i][1]))**4), 1)
```

We got the center of two by subtracting the same axis start coordinates and dividing them by two. In this way, the center of our detected rectangle is calculated. And at the end, a dot is drawn in the center. The default parameter for drawing boxes is a score of 0.5. if scores[0][i] >= 0.5 (i.e., equal or more than 50 percent) then we assume that the object is detected. if scores[0][i] >= 0.5:

```
mid x = (boxes[0][i][1] + boxes[0][i][3]) / 2
mid y = (boxes[0][i][0] + boxes[0][i][2]) / 2
apx distance = round(((1 - (boxes[0][i][3] -
boxes[0][i][1]))**4), 1)
```

In the above formula, mid_x is the center of the X-axis, and mid_y is the center of the y axis. If the distance apx_distance < 0.5 and if mid_x > 0.3 and mid_x < 0.7 then it can be concluded that the object is too close to the particular person. The relative distance between an object and a specific person can be computed using this code. Following the detection of an object, the algorithm is utilized to calculate the object's relative distance from the human. If the object is too close, then a signal or a warning is issued to the person through the voice generation module.

H. Voice Generation Module

After detecting an object, it is of utmost importance to acknowledge the person about the presence of that object on their way. For the voice generation module, PYTTTSX3 plays an important role. Pytttsx3 is a conversion library in Python which converts text into speech. This library works well with both Python 2 and 3. To get a reference to a pytttsx. Engine instance, a factory function called pytttsx.init. is invoked by an application. Pytttsx3 is a tool that converts text to speech easily.

This algorithm works whenever an object is detected and the approximate distance is being calculated. With the help of the cv2 library and cv2.putText() function, the texts are displayed on the screen. To identify the hidden text in an image, we use Python-tesseract for character recognition. OCR detects the text content on images and encodes it in a form easily understood by the computer. This text detection is done by scanning and analysis of the picture. Thus, the text embedded in images is recognized and "read" using Python-tesseract. Further, these texts are pointed to a pytttsx. Engine instance, a factory function called pytttsx.init. is invoked by an application. During construction, a pytttsx. Driver. DriverProxy object is initialized by an engine responsible for loading a speech engine driver from the pytttsx. driver's module. After construction, an entity created by a machine is used by the application to register and unregister event call-backs; produce and stop speech; get and set speech engine properties; and start and stop event loops.

Pytorch is primarily a machine learning library. Pytorch is mainly applied to the audio domain. Pytorch helps in loading the voice file in standard mp3 format. It also regulates the rate of audio dimension. Thus, it is used to manipulate the properties of sound like frequency, wavelength, and waveform. The numerous availabilities of options for audio synthesis can also be verified by looking at the functions of Pytorch.

IV. RESULTS AND DISCUSSION

Object detection is the emphasis of the proposed system. Wearable and portable technology has been developed. The person's chest is hooked to the system. The Raspberry Pi camera captures video of the scene, which is subsequently translated into frames by the processor. The auditory output from the system directs the user to the object. Figure 6 shows the detection of a person and a thing (blue cell phone). In Figure 7, a chair is seen with a person. The system displays the object's name its probability as a percentage. As a result, the system will only detect items with a probability more significant than the set threshold. Because You Only Look Once (YOLO) is employed to implement the design on the Android platform, the accuracy of object recognition is reduced. It shows the object's name as well as its probability. The program also informs the user of the class identification and the distance between the object and the camera over the device's speakers.

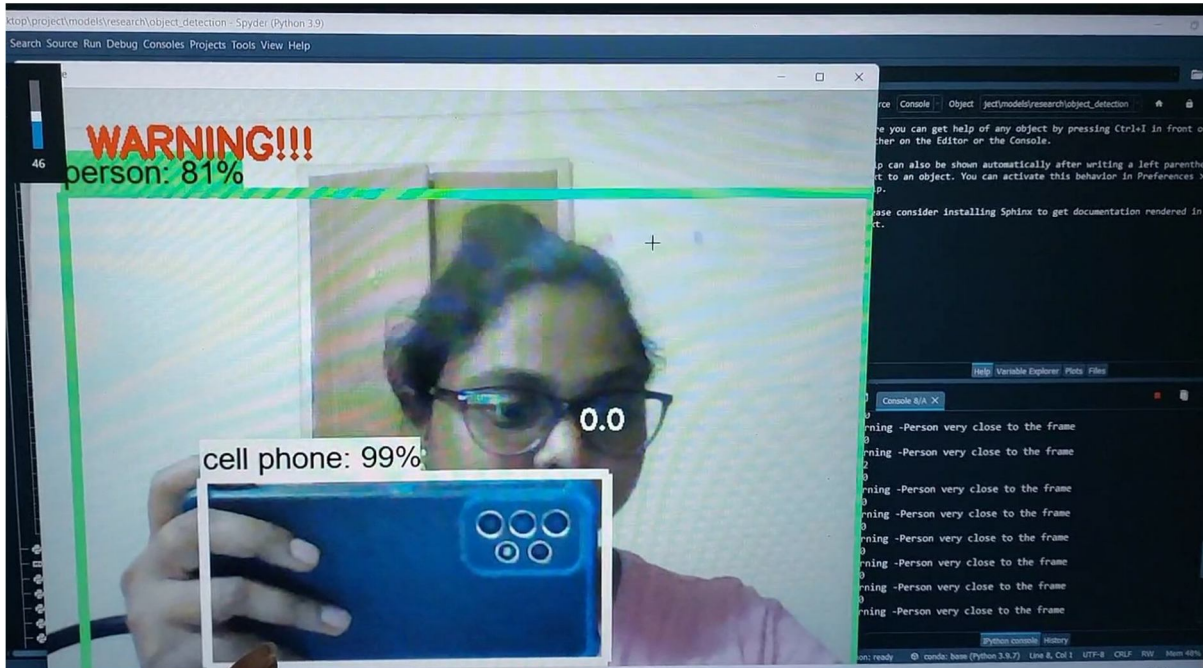


Fig 6. Image Recognition of person and cell phone

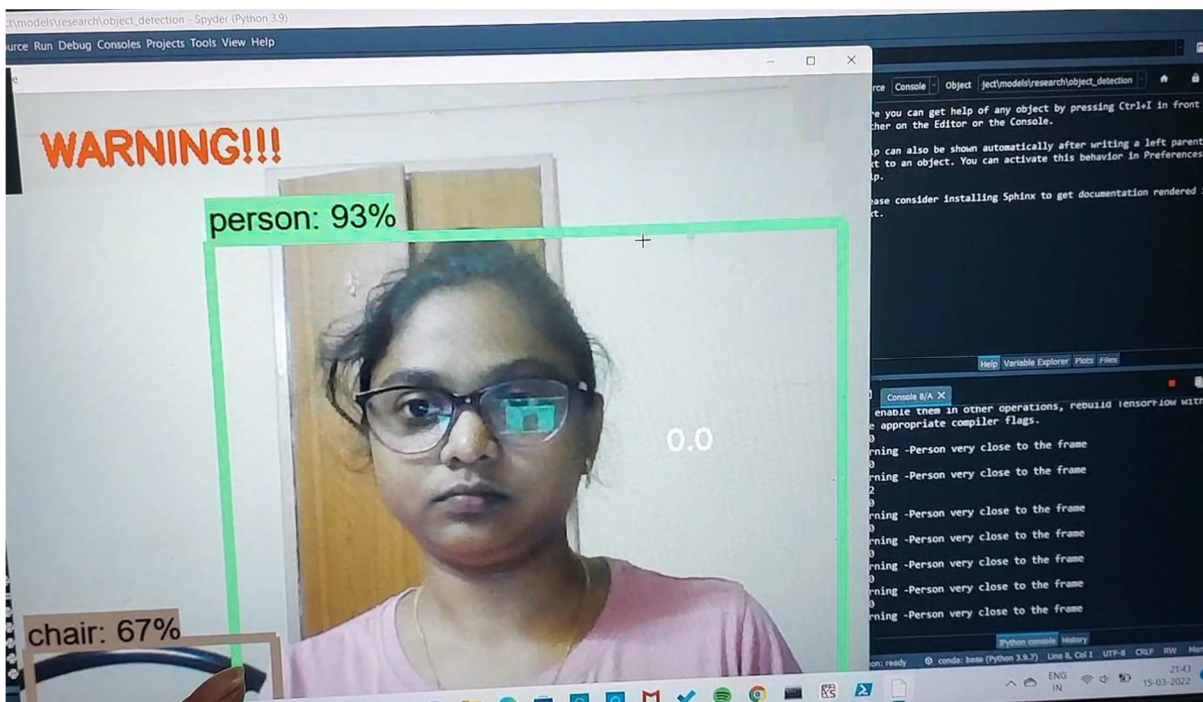


Fig 7. Image Recognition of person and chair

V. CONCLUSION

Several technologies have been created to aid visually impaired persons. One such attempt is that we would wish to make an Integrated Machine Learning System that allows the blind victims to identify and classify real-time objects generating voice feedback and distance. Which also produces warnings whether they are very close or far away from the thing. For visually blind folks, this technology gives voice direction. This technique has been introduced specifically to assist blind individuals. The precision, on the other hand, can be improved. Furthermore, the current system is based on the Android operating system; it can be altered to work with any device that is convenient.



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