ADVANCED AUTONOMOUS NAVIGATION AND ITS CONTROL ACTIONS

(MINOR PROJECT REPORT)

Submitted in
Fulfillment of the requirements for the degree of
Master of Technology
by

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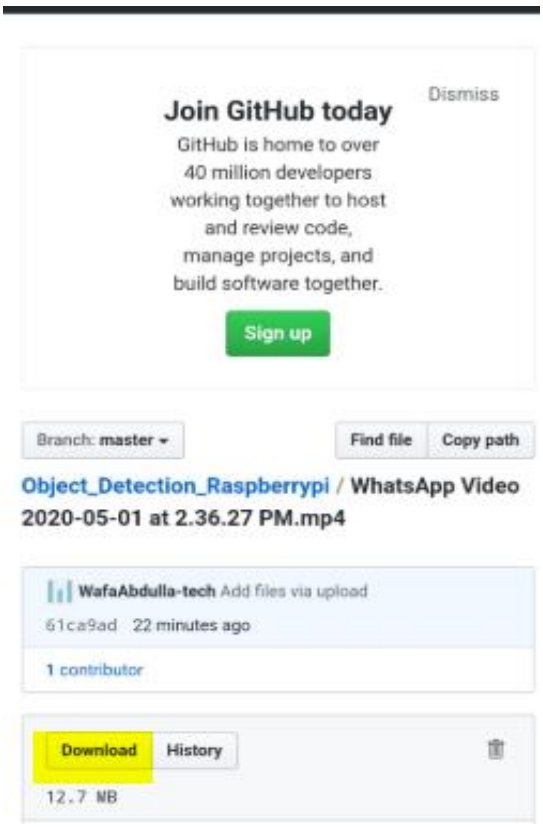
- Simulation results of object detection happened in laptop and webcam.
- Deployment of project in to Raspberrypi.
- Install All dependencies needs to run the project in Raspberrypi.
- Execute object detection project.
- Simulation Results of object_detection happened in Raspberrypi and Pi camera module.
- Real time Object detection using Raspberripi ,video available in below link.
- https://bit.ly/2xm67cC download it.



PART 3. control actions on speed performance using control system

- Future work

INTRODUCTION

Integrating vehicle autonomy requires system to have capability to perceive real world environment. In autonomous driving Neural Networks provides a substantial role in identifying obstacle, traffic signs etc.

This project has three stages of development. First stage involves interfacing camera with raspberry pi to capture real world images and differentiate it. Second stage involves enhancing with deep learning to identify and differentiate objects using trained Algorithm, and third stage performs control actions on speed, performance using control system which makes vital role in Autonomous Navigation.

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HARDWARE COMPONENTS :

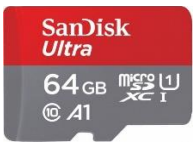
Specifications of Raspberry pi 3B+

The Raspberry Pi 3 Model B is the earliest model of the third-generation Raspberry Pi. It replaced the Raspberry Pi 2 Model B in February 2016. See also the Raspberry Pi 3 Model B+, the latest product in the Raspberry Pi 3 range.

- Quad Core 1.2GHz Broadcom BCM2837 64bit CPU
- 1GB RAM
- BCM43438 wireless LAN and Bluetooth Low Energy (BLE) on board
- 100 Base Ethernet
- 40-pin extended GPIO
- 4 Pole stereo output and composite video port
- Full size HDMI
- CSI camera port for connecting a Raspberry Pi camera
- DSI display port for connecting a Raspberry Pi touchscreen display
- Micro SD port for loading your operating system and storing data
- Upgraded switched Micro USB power source up to 2.5A

- Sd card

Secure Digital, officially abbreviated as **SD**, is a proprietary non-volatile **memory card** format developed by the **SD** Association (SDA) for use in portable devices



Pi Camera(5Mega Pixels-Version 1)

The **Pi camera** module is a portable light weight camera that supports Raspberry Pi. It communicates with Pi using the MIPI camera serial interface protocol. It is normally used in image processing, machine learning or in surveillance projects.

Ethernet Straight Cable

The **straight**-through is the most common type and is used to connect computers to hubs or switches.

PART -1

1.SET UP RASBERRYPI-3 AND INTERFACE CAMERA MODULE

- Download Rasbian buster lite from
  https://www.raspberrypi.org/downloads/raspbian/

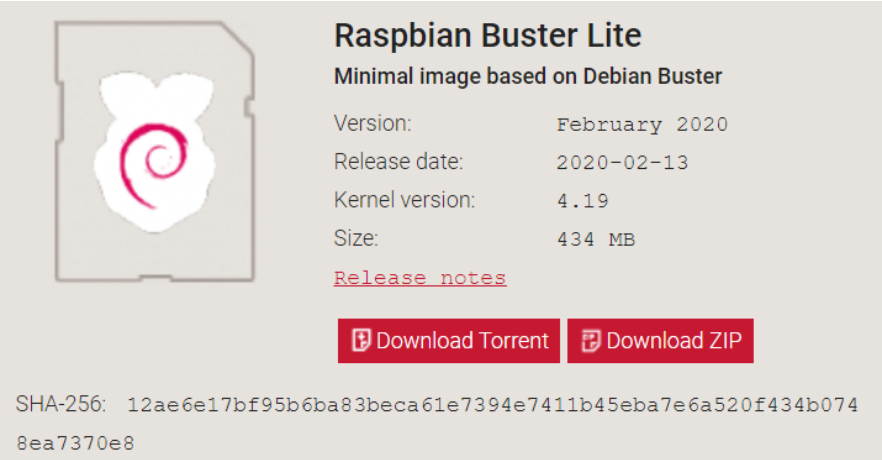


Fig 1.1 Rasbian Buster Lite

- Rasbian buster lite has pre installed os
- Put SD card in to laptop and format it using sd card formatter

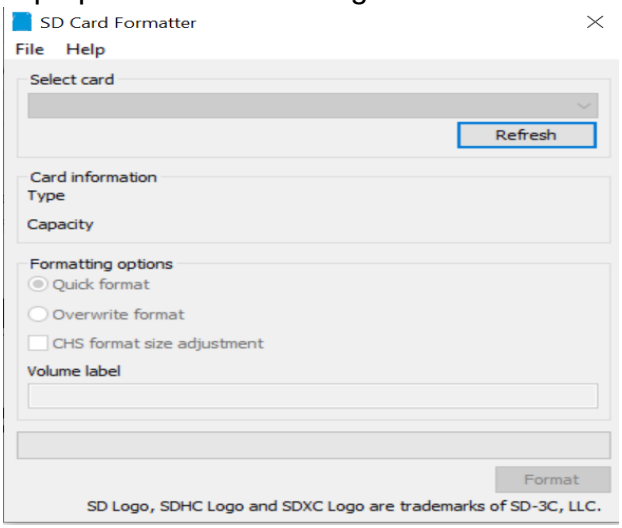


Fig 1.2 :sd card formatter

- Use Etcher balena software to flash the EEProm and to write the Rasbian buster lite .img file in to required sdcard.

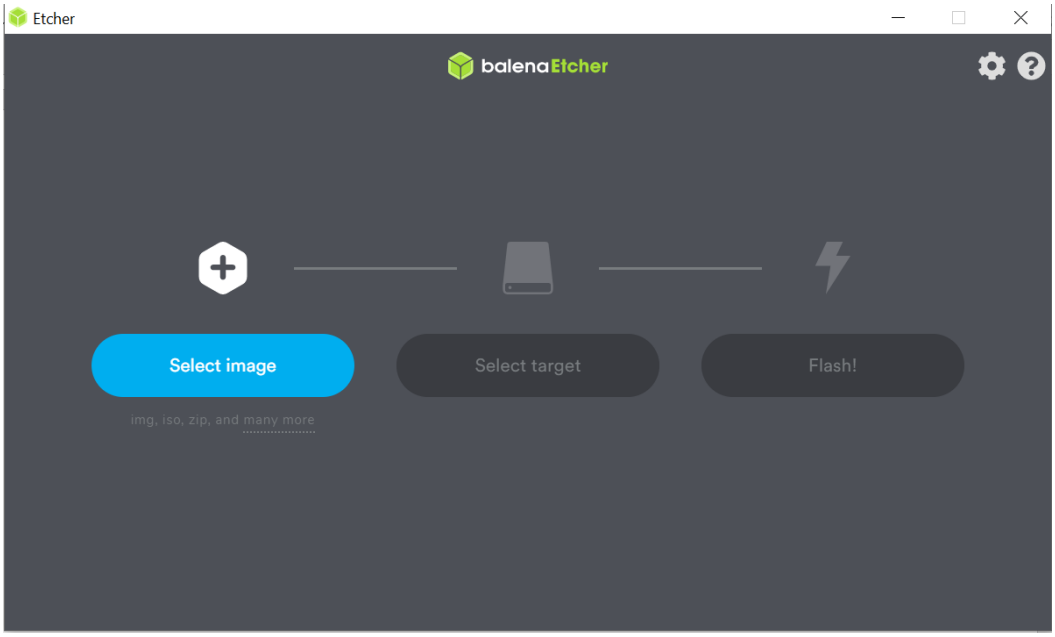


Fig 1.3:Etcher balena for to flash sd card

- Unmount the sdcard and put it again into laptop,open boot folder,add a file named ssh(without extension),Eject the sd card.
- Put the sd card in to Rasberrypi and plugin pi,connect an ethernet cable to Router.
- Connect wifi in laptop with the same network.
- Boot the sd card(ssh file will gets deleted from boot folder(which enables ssh in Rasberrypi)
- Check the ip address of Rasberrypi.

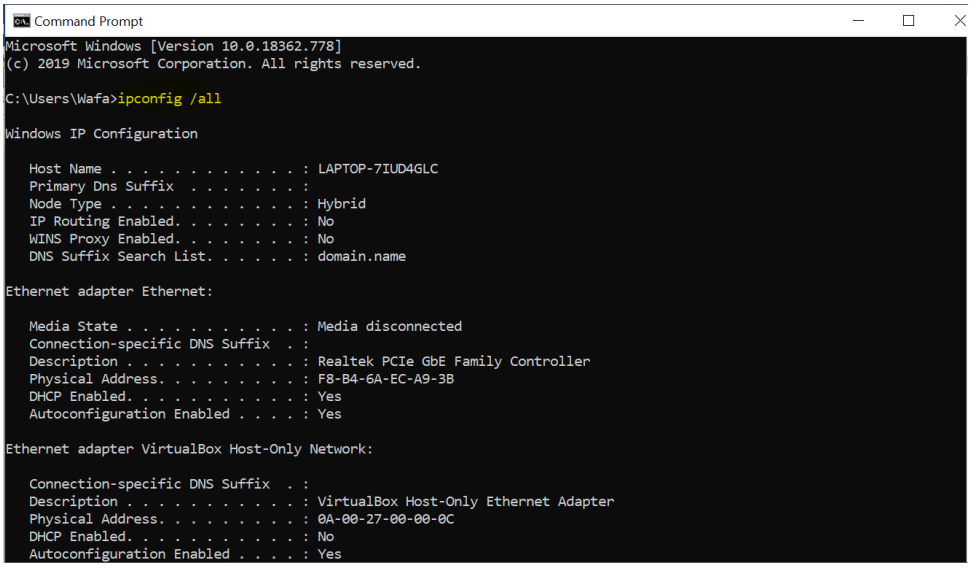


Fig 1.4:command prompt,ipconfig /all

- Get the ip address of router

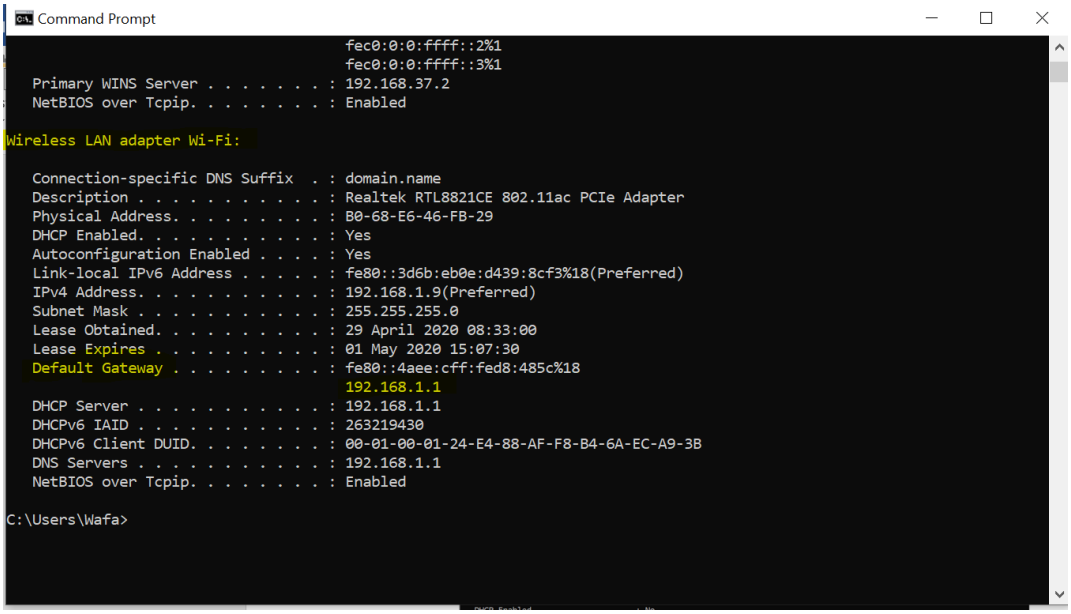


Fig 1.5:Get the gateway address of wifi

- Check that ip address in google,In Router page check the DHCP list and get ip address of Rasberrypi(it will be in the range (192.168.1.(1-254)

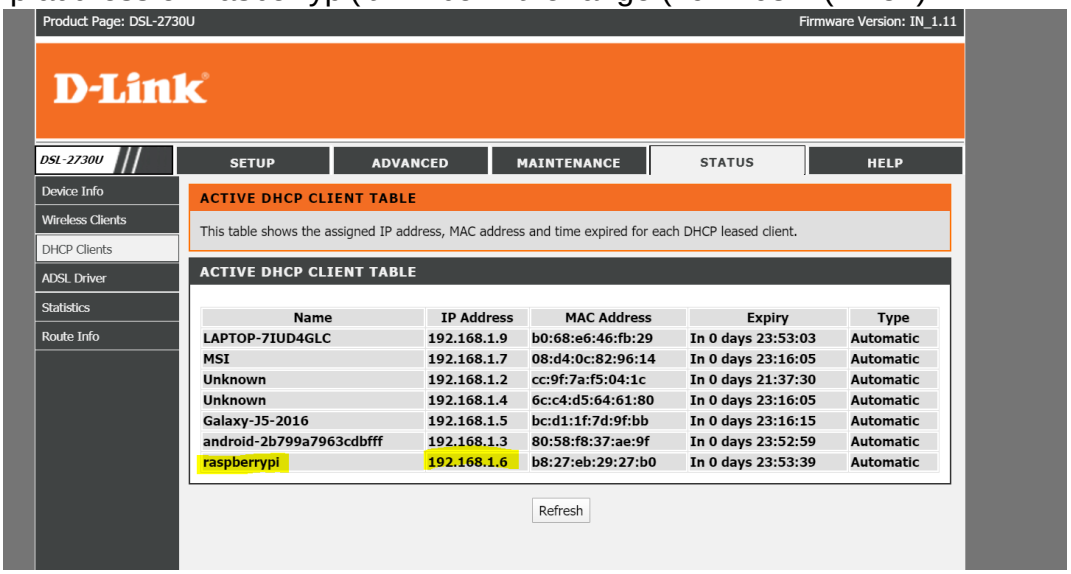


Fig 1.6:Get ip address of Raspberrypi from routers DHCP client list

- Assign this ip address as static ip address of Rasberry pi.
- Remove the SD card from Rasberrypi,and connect to laptop using card reader,open boot folder,open cmdline.txt,append ip=192.168.1.6 at the end and save it.
- Remove sd card and put it back to Rasberrypi ,boot it.
- Open ssh(putty),type host address as the static ip,login as username pi ans password raspberry as default.

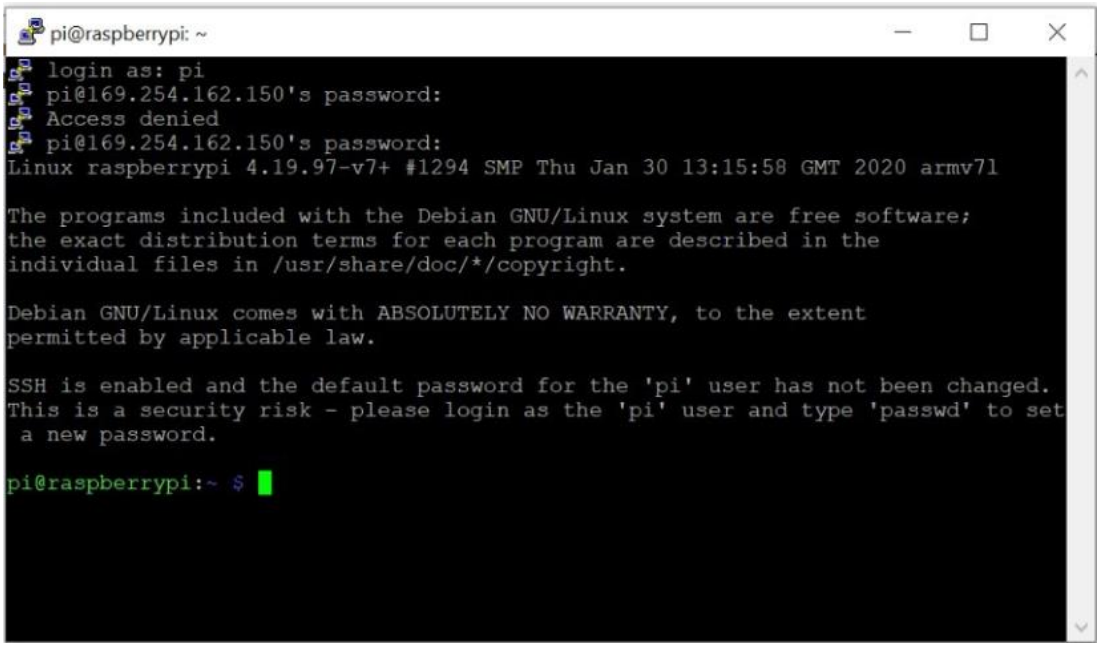


Fig 1.7:Login Raspberrypi through Putty

- Enable vnc server,type sudo raspi-config(go to interfaces,enable vnc server).
- sudo apt-get install lxsession execute the command for Debain GUI Desktop.(In boot option,select Desktop)and Reboot Raspberrypi.
- Install vnc viewer

Fig 1.8:Connect to vnc viewer using Raspberrypis ip address

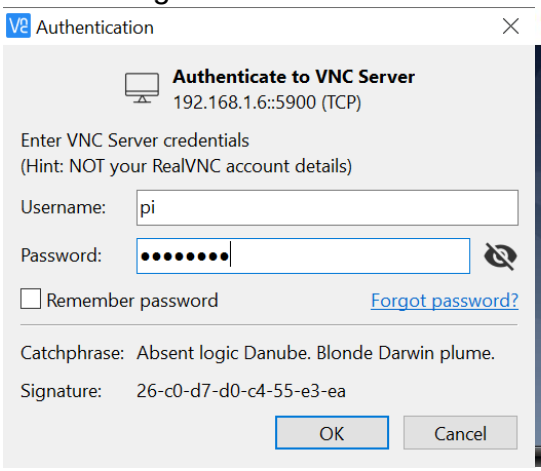


Fig 1.9:Login to VNC viewer

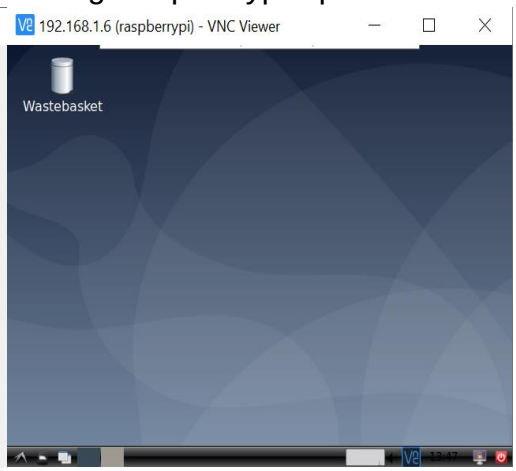


Fig 1.10:Debian desktop,through VNC viewer

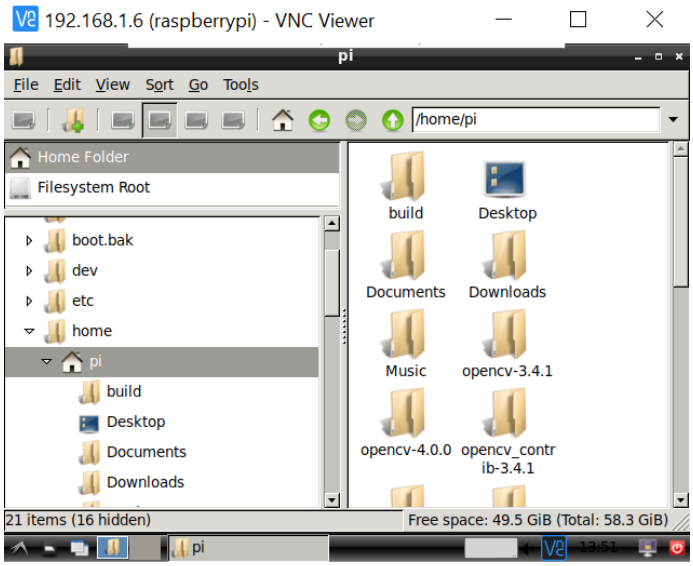


Fig 1.11:pi directory

CAMERA INTERFACING:

Fig 1.12:Camera interfacing

.Enable camera module in interfaces and Reboot Rasberrypi.

Fig 1.13:Camera Initialization

- Test the camera module by capture an image

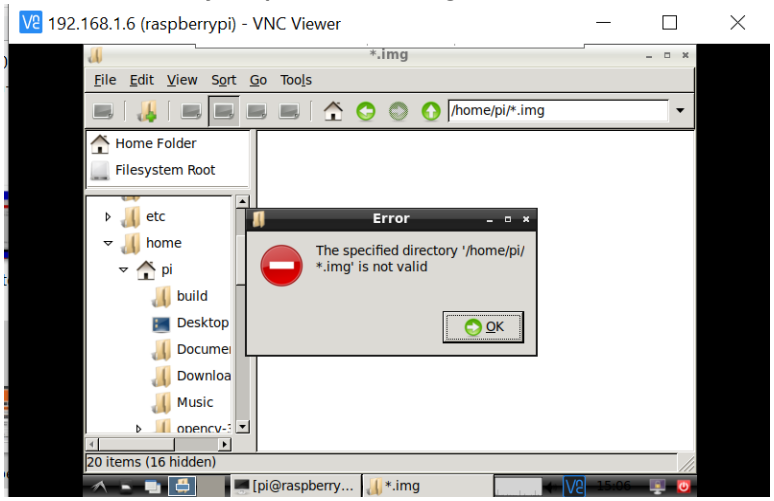


Fig 1.14:a.Testing camera module

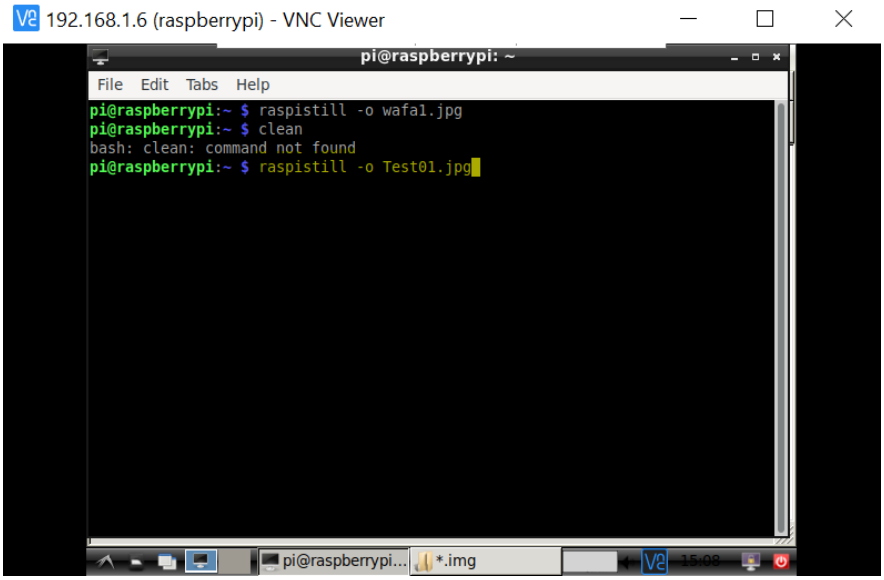


Fig 1.14:b.Testing camera module

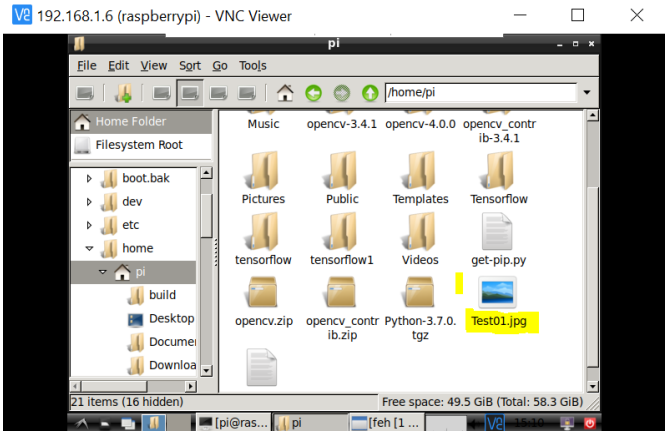


Fig 1.14:c.Testing camera module

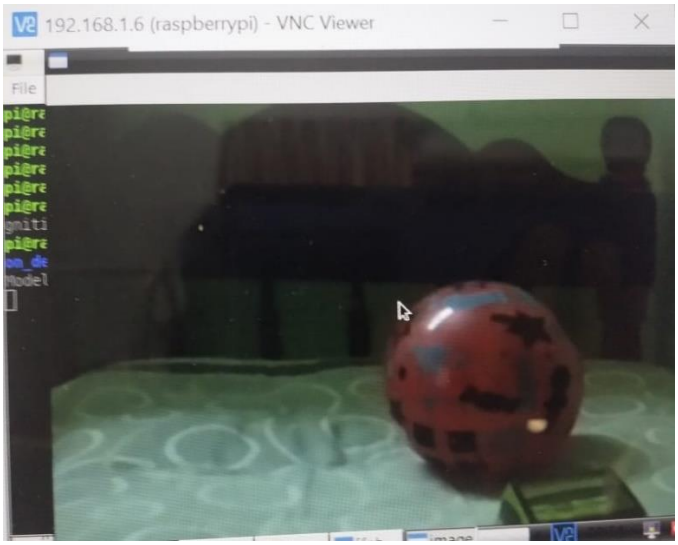


Fig 1.15:Captured image

PART-2

Enhancing with deep learning to identify and differentiate objects using trained Algorithm

- Testing of tensor flow object detection API in laptop using webcam(to capture real time image).

TENSOR FLOW OBJECT DETECTION API:

Dependencies

Tensorflow Object Detection API depends on the following libraries:

- Protobuf 3.0.0
- Python-tk
- Pillow 1.0
- lxml
- tf Slim (which is included in the "tensorflow/models/research/" checkout)
- Jupyter notebook
- Matplotlib
- Tensorflow (>=1.12.0)
- Cython
- contextlib2

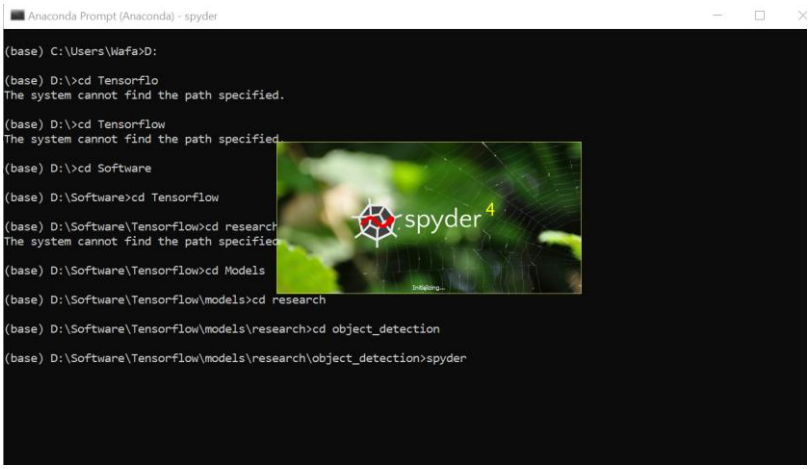
- cocoapi

Protobuf Compilation

The Tensorflow Object Detection API uses Protobufs to configure model and training parameters. Before the framework can be used, the Protobuf libraries must be compiled.

Install Anaconda for Windows, Open Anaconda prompt,Type spyder or jupiter notebook**,** Create a new project in spyder,and create the .py file,run section by section

---

- 

. Fig 2.1:Open spyder in Anaconda

Create new project and execute the code

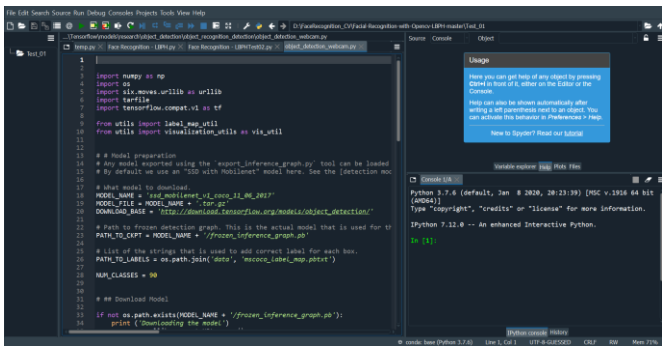


Fig 2.2:Create new python project in spyder

Run the file

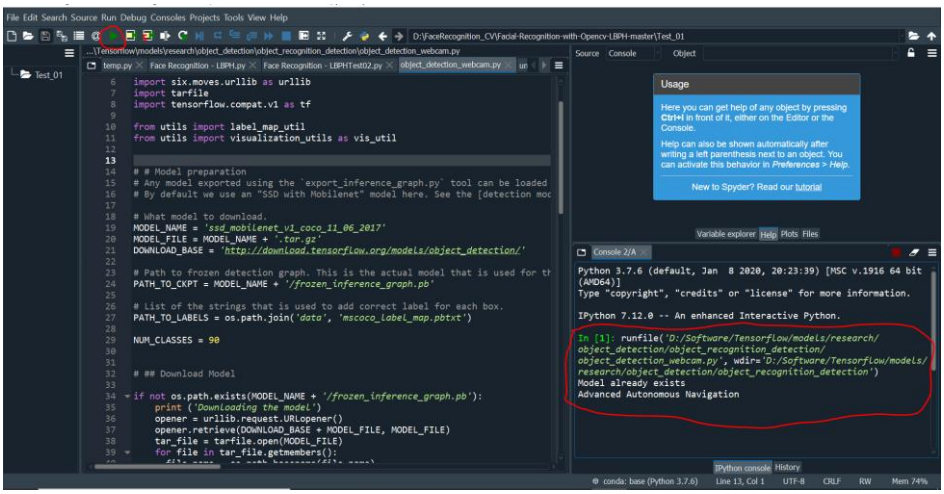


Fig 2.3:Run Object_detection program

- **Object_detection_webcam.py file(Program)**

import numpy as np

import os

import six.moves.urllib as urllib

import sys

import tarfile

import tensorflow.compat.v1 as tf

import zipfile

from collections import defaultdict

from io import StringIO

from matplotlib import pyplot as plt

from PIL import Image

from utils import label_map_util

from utils import visualization_utils as vis_util

Model preparation

```
# Any model exported using the `export_inference_graph.py` tool can be loaded here
simply by changing `PATH_TO_CKPT` to point to a new .pb file.

# By default we use an "SSD with Mobilenet" model here. See the [detection model
zoo](https://github.com/tensorflow/models/blob/master/object_detection/g3doc/detecti
on_model_zoo.md) for a list of other models that can be run out-of-the-box with
varying speeds and accuracies.



 # What model to download.

MODEL_NAME = 'ssd_mobilenet_v1_coco_11_06_2017'

MODEL_FILE = MODEL_NAME + '.tar.gz'

DOWNLOAD_BASE = 'http://download.tensorflow.org/models/object_detection/'

# Path to frozen detection graph. This is the actual model that is used for the object
detection.

PATH_TO_CKPT = MODEL_NAME + '/frozen_inference_graph.pb'

# List of the strings that is used to add correct label for each box.

PATH_TO_LABELS = os.path.join('data', 'mscoco_label_map.pbtxt')

NUM_CLASSES = 90

# ## Download Model

if not os.path.exists(MODEL_NAME + '/frozen_inference_graph.pb'):

        print ('Downloading the model')

        opener = urllib.request.URLopener()

        opener.retrieve(DOWNLOAD_BASE + MODEL_FILE, MODEL_FILE)

        tar_file = tarfile.open(MODEL_FILE)

        for file in tar_file.getmembers():

          file_name = os.path.basename(file.name)

          if 'frozen_inference_graph.pb' in file_name:

            tar_file.extract(file, os.getcwd())

        print ('Download complete')

else:
```

```
                print ('Model already exists')


         # ## Load a (frozen) Tensorflow model into memory.

         detection_graph = tf.Graph()

        with detection_graph.as_default():

        od_graph_def = tf.GraphDef()

        with tf.gfile.GFile(PATH_TO_CKPT, 'rb') as fid:

        serialized_graph = fid.read()

        od_graph_def.ParseFromString(serialized_graph)

        tf.import_graph_def(od_graph_def, name='')

        # ## Loading label map

        # Label maps map indices to category names, so that when our convolution network predicts `5`, we know that this corresponds to `airplane`.  Here we use internal utility functions, but anything that returns a dictionary mapping integers to appropriate string labels would be fine

       label_map = label_map_util.load_labelmap(PATH_TO_LABELS)

       categories = label_map_util.convert_label_map_to_categories(label_map, max_num_classes=NUM_CLASSES, use_display_name=True)

          category_index = label_map_util.create_category_index(categories)

       #intializing the web camera device

       import cv2

      cap = cv2.VideoCapture(0)

      # Running the tensorflow session

      with detection_graph.as_default():

            with tf.Session(graph=detection_graph) as sess:

            ret = True

            while (ret):

              ret,image_np = cap.read()
```

```
      # Expand dimensions since the model expects images to have shape: [1, None,
None, 3]

      image_np_expanded = np.expand_dims(image_np, axis=0)

      image_tensor = detection_graph.get_tensor_by_name('image_tensor:0')

      # Each box represents a part of the image where a particular object was
detected.

      boxes = detection_graph.get_tensor_by_name('detection_boxes:0')

      # Each score represent how level of confidence for each of the objects.

      # Score is shown on the result image, together with the class label.

      scores = detection_graph.get_tensor_by_name('detection_scores:0')

      classes = detection_graph.get_tensor_by_name('detection_classes:0')

      num_detections = detection_graph.get_tensor_by_name('num_detections:0')

      # Actual detection.

      (boxes, scores, classes, num_detections) = sess.run(

          [boxes, scores, classes, num_detections],

          feed_dict={image_tensor: image_np_expanded})

      # Visualization of the results of a detection.

      vis_util.visualize_boxes_and_labels_on_image_array(

          image_np,

          np.squeeze(boxes),

          np.squeeze(classes).astype(np.int32),

          np.squeeze(scores),

          category_index,

          use_normalized_coordinates=True,

          line_thickness=8)

#       plt.figure(figsize=IMAGE_SIZE)

#       plt.imshow(image_np)
```

```
cv2.imshow('image',cv2.resize(image_np,(1280,960)))
if cv2.waitKey(25) & 0xFF == ord('q'):
    cv2.destroyAllWindows()
    cap.release()
    break
```

Results(Laptop)

Fig 2.4:Simulation Result laptop(Detecting car)

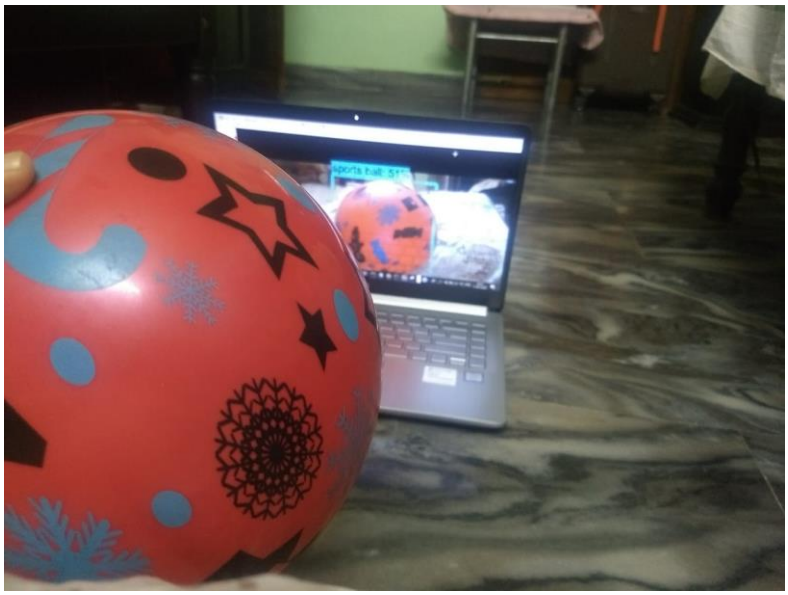


Fig 2.5:Simulation Result laptop(Detecting sports ball)

- **Deployment of the project in to Raspberrypi**
- Use Winscp for file transfer

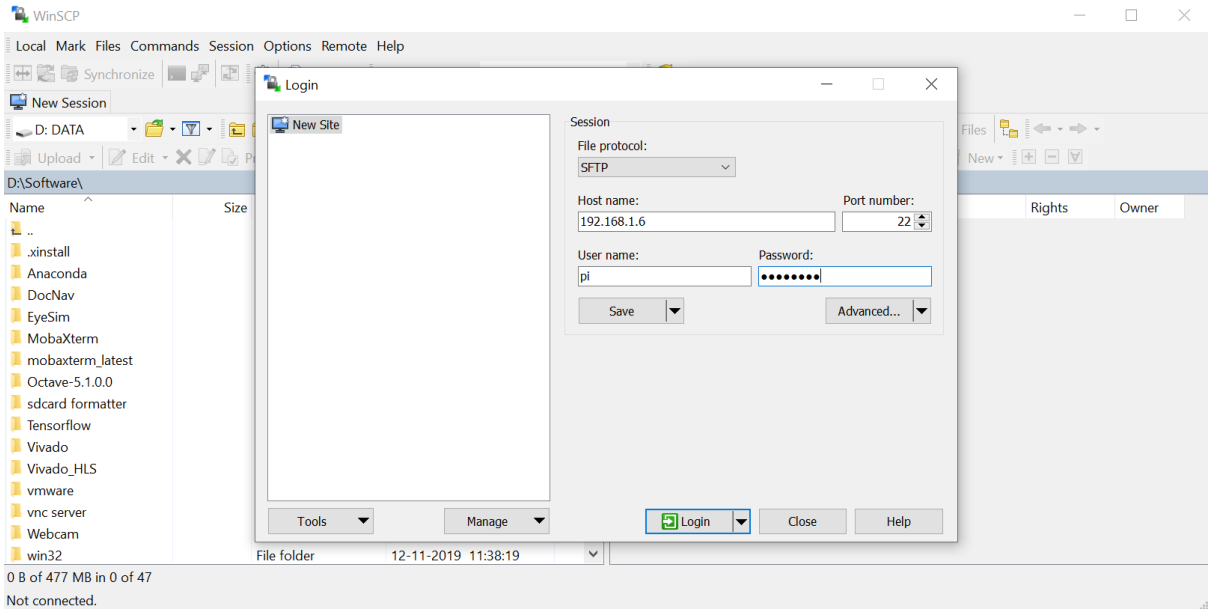


Fig 2.6:Login winscp using RaspberryPi's ip and credaintails

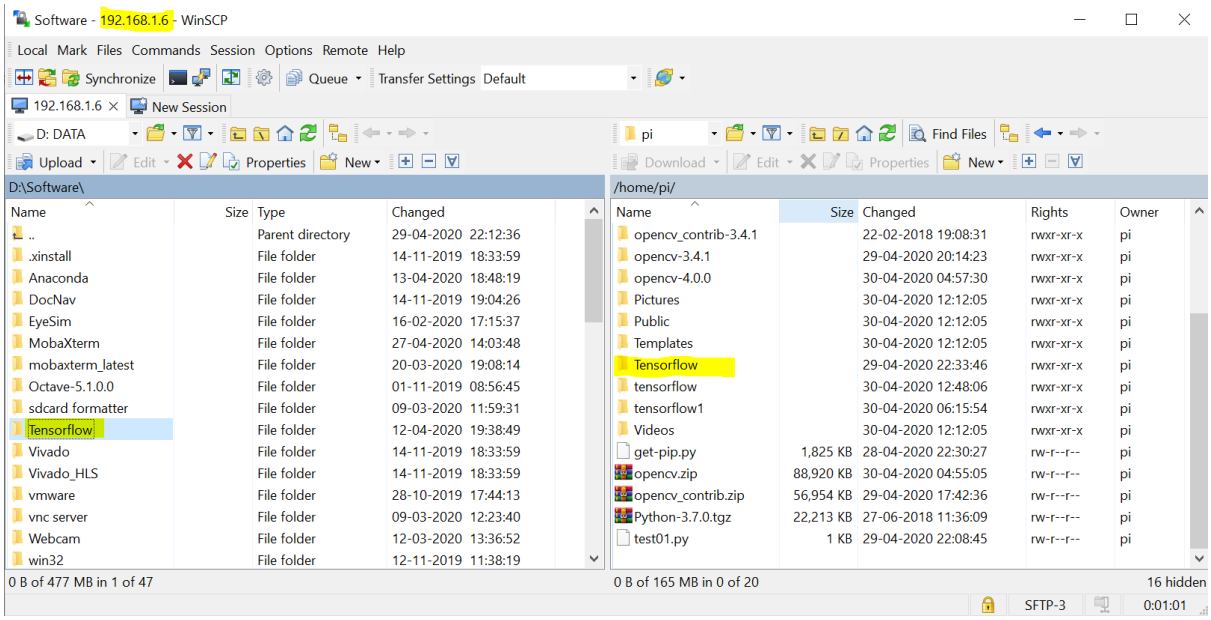


Fig 2.7:copy files from laptop to Raspberry

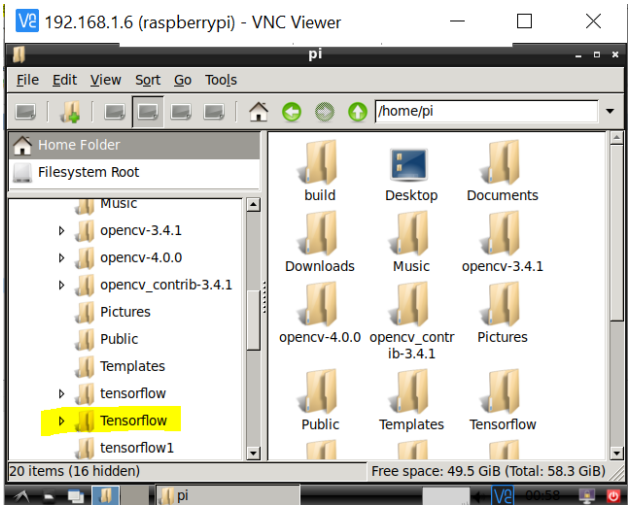


Fig 2.8:Verify files Deployed properlyin Raspberry

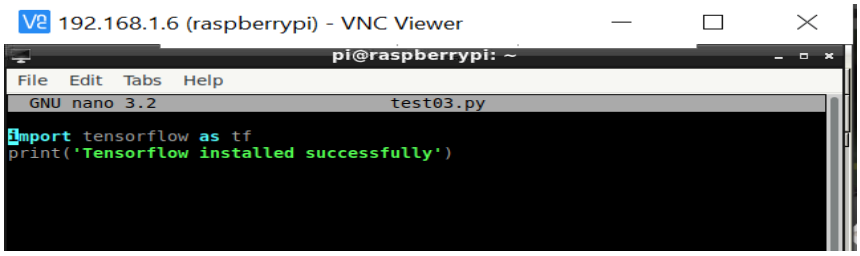


Fig 2.9:a.Verify Tensorflow installed properly

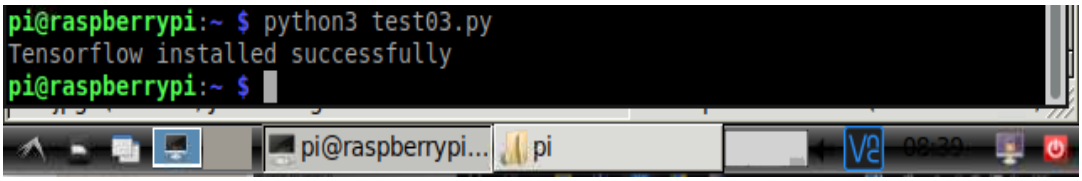


Fig 2.9:b.Verify Tensorflow installed properly

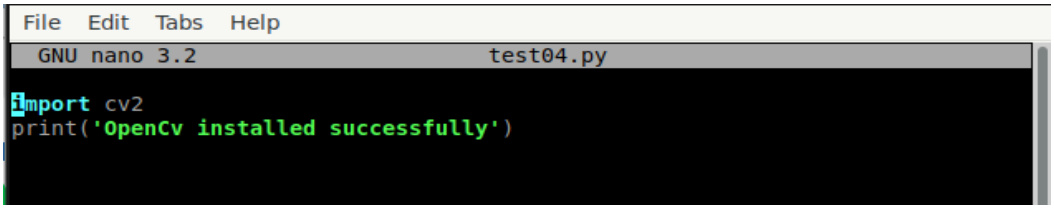


Fig 2.10:a.Verify OpenCVinstalled properly

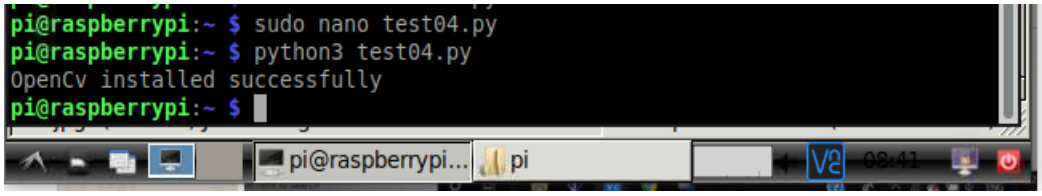


Fig 2.10:b.Verify OpenCVinstalled properly

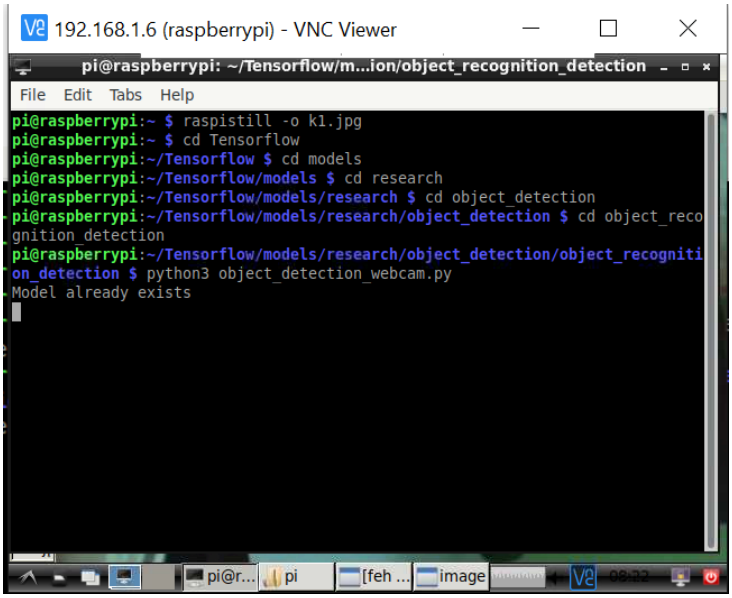


Fig 2.11:Execute Object_detection program

RESULTS(RASPBERRY PI)

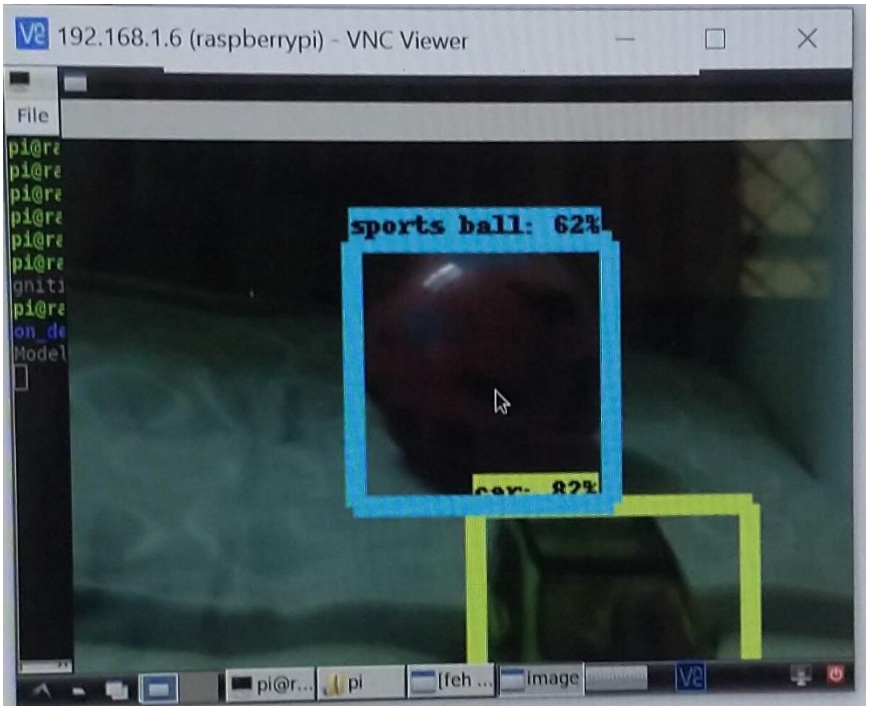


Fig 2.12:Simulation result(Raspberry)-1

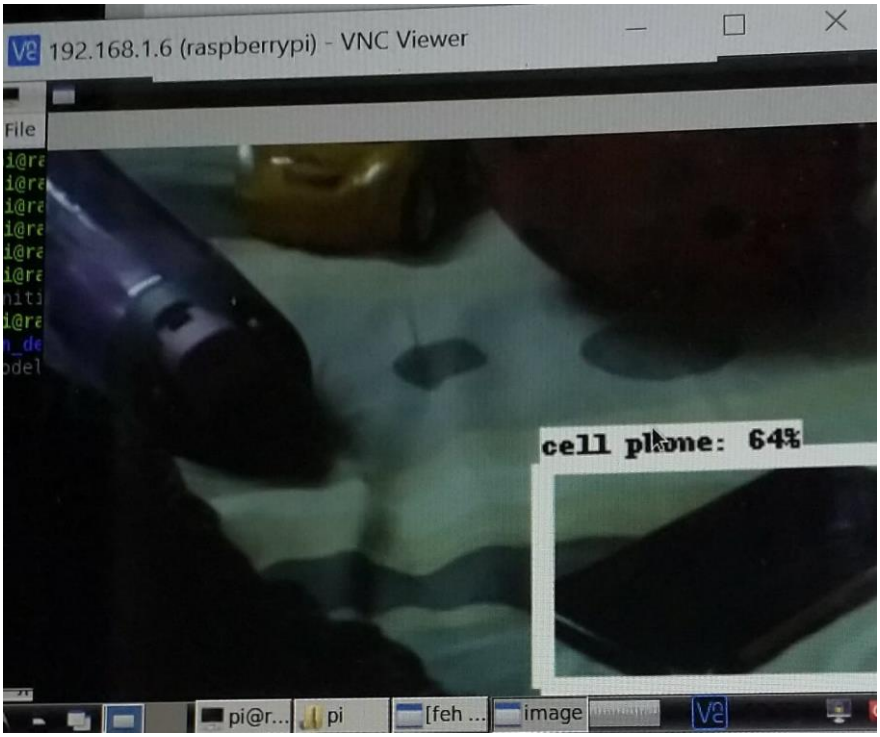


Fig 2.13:Simulation result(Raspberry)-2

CONCLUSION

- Done set up of Rasberrypi and interfaced camera module
- Achieved Object detection in laptop , deployed in Raspberrypi and Tested.

References:

- https://ieeexplore.ieee.org/document/8494053
- https://ieeexplore.ieee.org/document/9070240
- https://ieeexplore.ieee.org/document/8452728
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- https://www.datacamp.com/community/tutorials/tensorflow-tutorial
- https://projects.raspberrypi.org/en/projects/getting-started-with-picamera/5
- https://www.raspberrypi.org/documentation/remote-access/vnc/
- Install Rasbian Buster - https://www.youtube.com/watch?v=AuvbbP7Dues
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- OpenCV Installation - https://pysource.com/2018/10/31/raspberry-pi-3-and-opencv-3-installation-tutorial/
- https://www.analyticsvidhya.com/blog/2016/10/an-introduction-to-implementing-neural-networks-using-tensorflow/
- https://www.youtube.com/watch?v=MoMjIwGSFVQ
- https://www.analyticsvidhya.com/blog/2018/06/understanding-building-object-detection-model-python/
- NOOBS Installation - https://www.youtube.com/watch?v=wvxCNQ5AYPg
- Static IP setup - https://www.youtube.com/watch?v=3VAlvtFWOwE