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# Autonomous Drone Follow

— Unmanned Aerial Vehicles —

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# Task

- Accurate depth calculation
- Minimum Computation
- Maximum FPS

# Literature Survey

## Measuring Distance with Mobile Phones Using Single-Camera Stereo Vision

**Abstract**—Computer stereo vision is an important technique for robotic navigation and other mobile scenarios where depth perception is needed, but it usually requires two cameras with a known horizontal displacement. In this paper, we present a solution for mobile devices with just one camera, which is a first step towards making computer stereo vision available to a wide range of devices that are not equipped with stereo cameras. We have built a prototype using a state-of-the-art mobile phone, which has to be manually displaced in order to record images from different lines of sight. Since the displacement between the two images is not known in advance, it is measured using the phone's inertial sensors. We evaluated the accuracy of our single-camera approach by performing distance calculations to everyday objects in different indoor and outdoor scenarios, and compared the results with that of a stereo camera phone. As a main advantage of a single moving camera is the possibility to vary its relative position between taking the two pictures, we investigated the effect of different camera displacements on the accuracy of distance measurements.

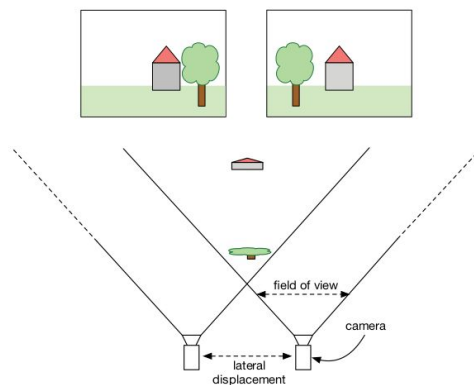
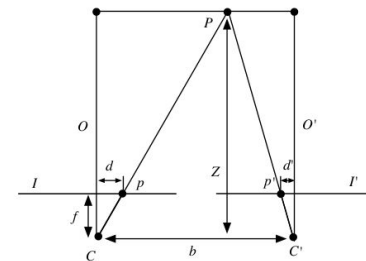
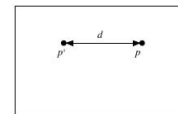


Figure 1. Two images recorded from a laterally displaced camera. The displacement of the objects provides relative depth information.



(a) Geometry of stereo triangulation.



(b) Overlay of the images shows the difference between the image locations  $p$  and  $p'$ .

# Maths Used

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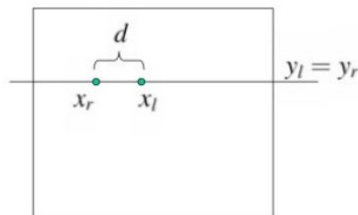
## Recall: Stereo Disparity

Left camera

$$x_l = f \frac{X}{Z} \quad y_l = f \frac{Y}{Z}$$

Right camera

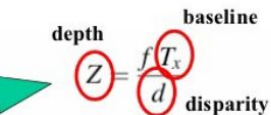
$$x_r = f \frac{X - T_x}{Z} \quad y_r = f \frac{Y}{Z}$$



Stereo Disparity

$$d = x_l - x_r = f \frac{X}{Z} - (f \frac{X}{Z} - f \frac{T_x}{Z})$$

$$d = \frac{f T_x}{Z}$$



**Important equation!**

## Triangle Similarity for Object/Marker to Camera Distance

In order to determine the distance from our camera to a known object or marker, we are going to utilize *triangle similarity*.

As I continue to move my camera both closer and farther away from the object/marker, I can apply the triangle similarity to determine the distance of the object to the camera:

$$D' = (W \times F) / P$$

### 1.1 What is the range of the Depth measured by the camera?

Depth Range is from **50 cm to 300cm**. The maximum working distance varies with the lens. With our default lens, the maximum working distance is about 300cm, beyond which the depth accuracy drops down

# Approach

- Object detection with single camera
- Stereo vision camera for depth and width calculation
- Size based depth calculation (if object width known)

```
def depth():
    for i in range(len(files)):
        imgR = cv2.imread(right + files[i])
        bboxR, labelR, confR = cv.detect_common_objects(imgR)
        if len(bboxR)>0 and labelR[0] == 'person' and flag==0:
            a = int(files[i][5:-4])
            a = a-1
            a = str(a)
            a = '{0}'.format(a.zfill(4))
            imgL = cv2.imread(left + files[i][0:5] + a + ".jpg")
            bboxL, labelL, confL = cv.detect_common_objects(imgL)

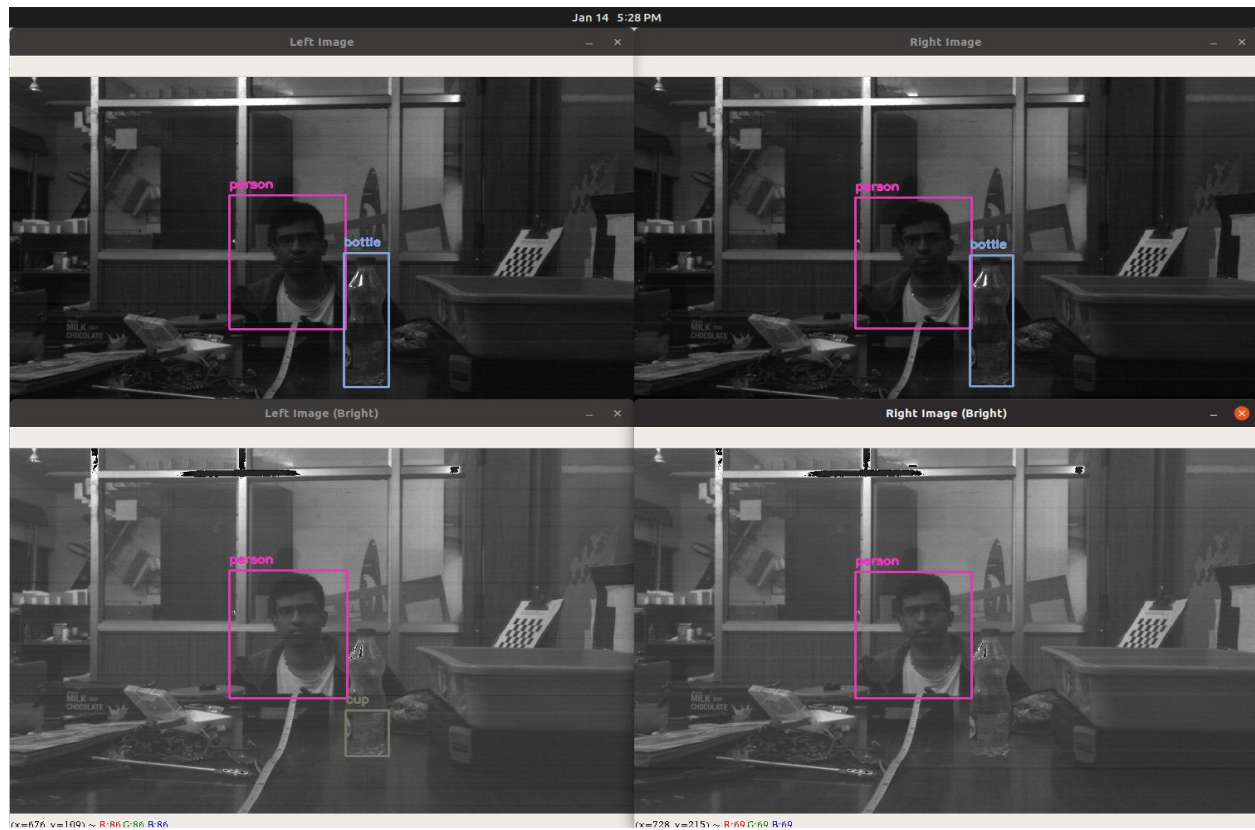
            if len(bboxL) == 0:
                print("Object only in Right Camera")
            else:
                x0 = int((bboxR[0][0]+bboxR[0][2])/2)
                x1 = int((bboxL[0][0]+bboxL[0][2])/2)
                d = abs(x1-x0)
                z = (f*baseline)/d
                print(z)

                p = bboxR[0][2] - bboxR[0][0]
                w = (z*p)/f
                flag=1

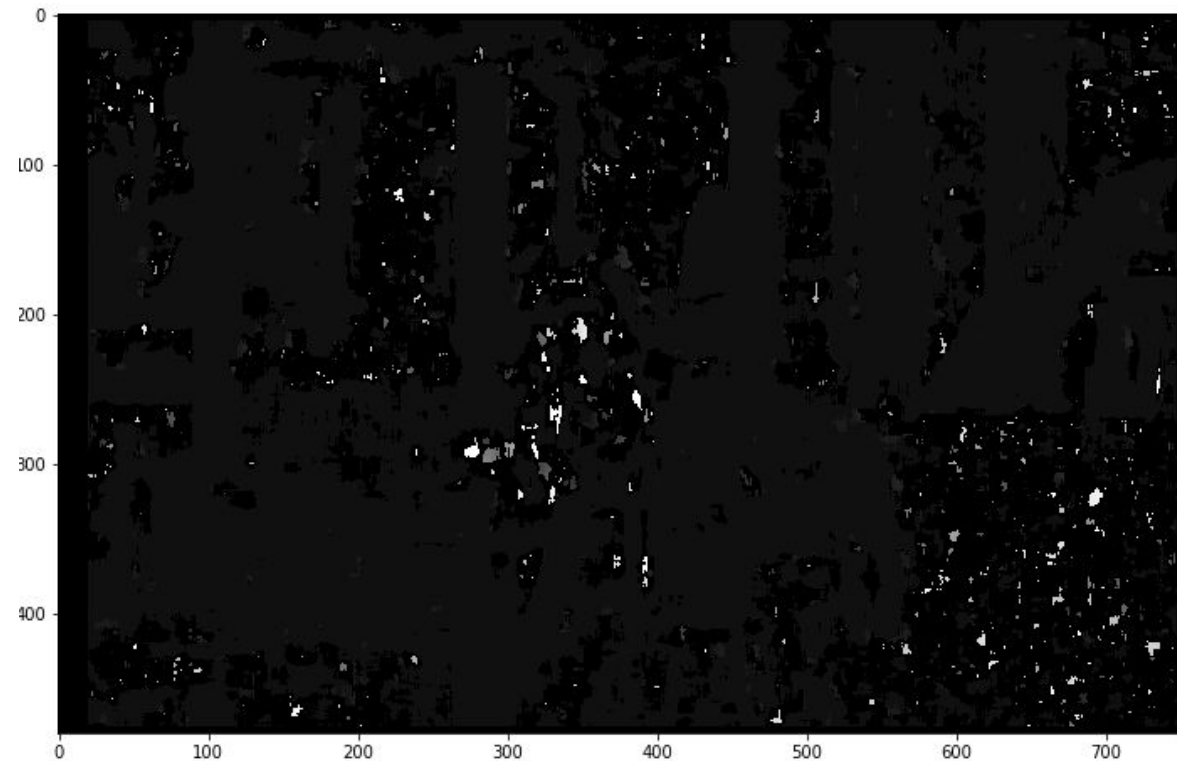
        elif len(bboxR)>0 and labelR[0] == 'person' and flag==1:
            p = bboxR[0][2] - bboxR[0][0]
            z = (w*f)/p
            print(z)

        else:
            print("No Object")
            flag=0
```

# Experiments



# Experiments

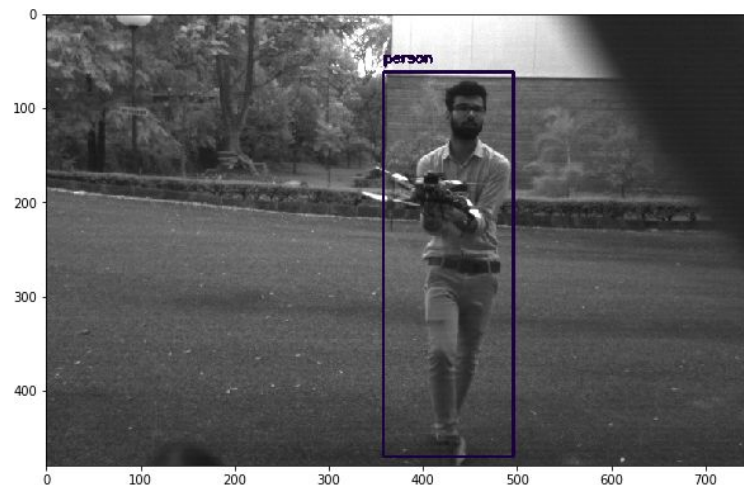
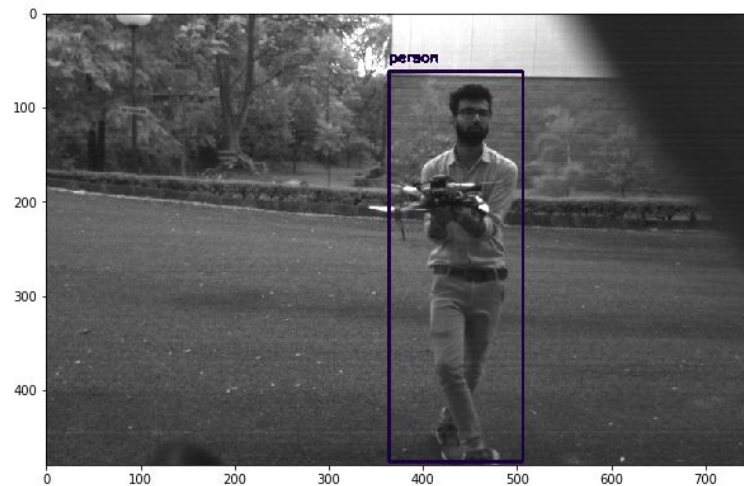


```
array([[ -16,  -16,  -16, ...,  -16,  -16,  -16],  
       [ -16,  -16,  -16, ...,  -16,  -16,  -16],  
       [ -16,  -16,  -16, ...,  -16,  -16,  -16],  
       ...,  
       [ -16,  -16,  -16, ...,  -16,  -16,  -16],  
       [ -16,  -16,  -16, ...,  -16,  -16,  -16],  
       [ -16,  -16,  -16, ...,  -16,  -16,  -16]], dtype=int16)
```



# Experiments

```
No Object
No Object
No Object
Object only in Right Camera
No Object
2533.3333333333335
3065.193370165746
2859.7938144329896
3263.529411764706
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
No Object
Object only in Right Camera
No Object
4560.0
5461.077844311377
5922.077922077922
5629.62962962963
5211.428571428572
7238.0952380952385
5461.077844311377
6656.934306569343
6246.575342465753
6561.151079136691
6857.142857142857
6961.832061068702
```





# Experiments

```
... ..  
No Object  
No Object  
No Object  
1543.1472081218274  
No Object  
No Object  
1717.5141242937852  
2303.030303030303  
2082.1917808219177  
2268.6567164179105  
2320.610687022901  
3070.7070707070707  
2980.392156862745  
2320.610687022901  
2375.0  
2303.030303030303  
2235.294117647059  
2980.392156862745  
2788.9908256880735  
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2285.714285714286  
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2000.0  
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2356.5891472868216  
2140.845070422535  
1974.025974025974  
2026.6666666666667  
1727.2727272727273  
1608.4656084656085  
1551.0204081632653  
1375.5656108597284  
No Object  
No Object  
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No Object  
1256.198347107438  
No Object  
1388.1278538812785  
1679.5580110497237  
1567.0103092783504  
1788.235294117647
```

# Results

- Using CVLIB library for object detection
  - Object detection in single frame : 0.20 (FPS)
  - Depth calculation : 0.85 (FPS)
- Using Tiny Yolo weights
  - Object detection in single frame : 6.74 (FPS)
  - Depth calculation : 8.42 (FPS)

**Result's are of CPU and experimental based.**

# Problem

- Depth is not accurate (dependency on disparity)
  - Bounding boxes along the object is not perfect – due to which disparity is not accurate
- Solution (Future Work)
  - 1st method
    - Average Disparity
  - 2nd method
    - Instead of taking point's from bounding box need to take other method.
    - Trying to use concept of template matching, reading SURF and ORB techniques to get accurate results

# Experiments

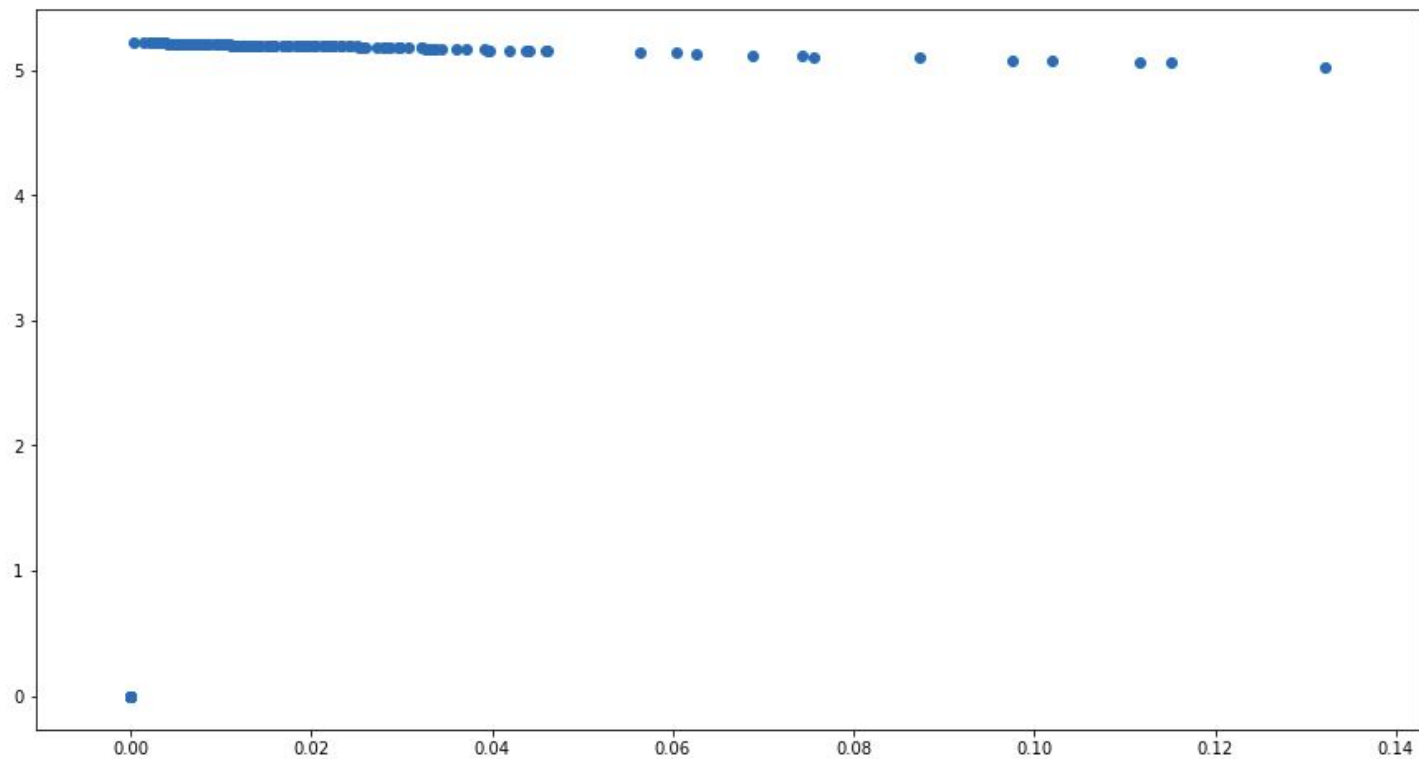
## ORB Detector

- Disparity based on descriptor's shift

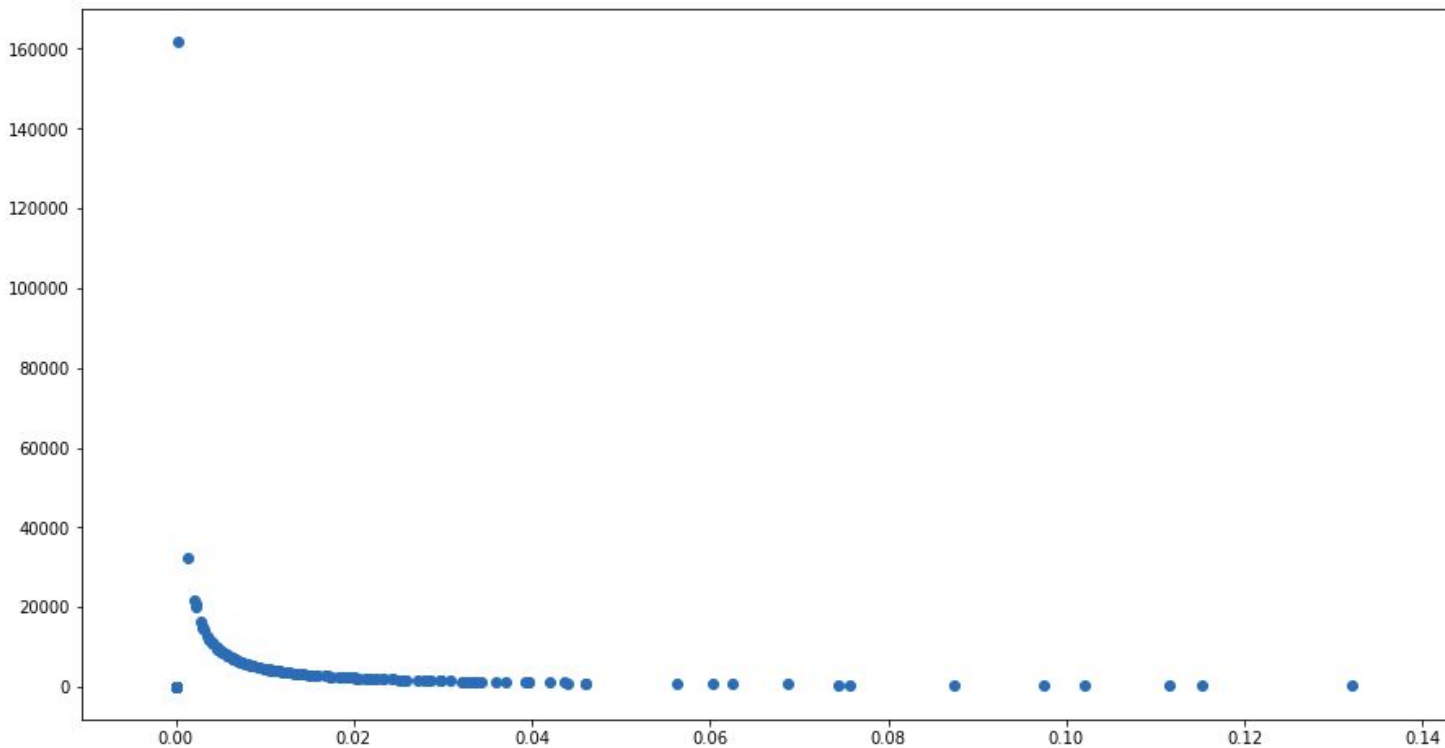
## Average Disparity

- Average Disparity
  - Curve fitting
  - Using Formula

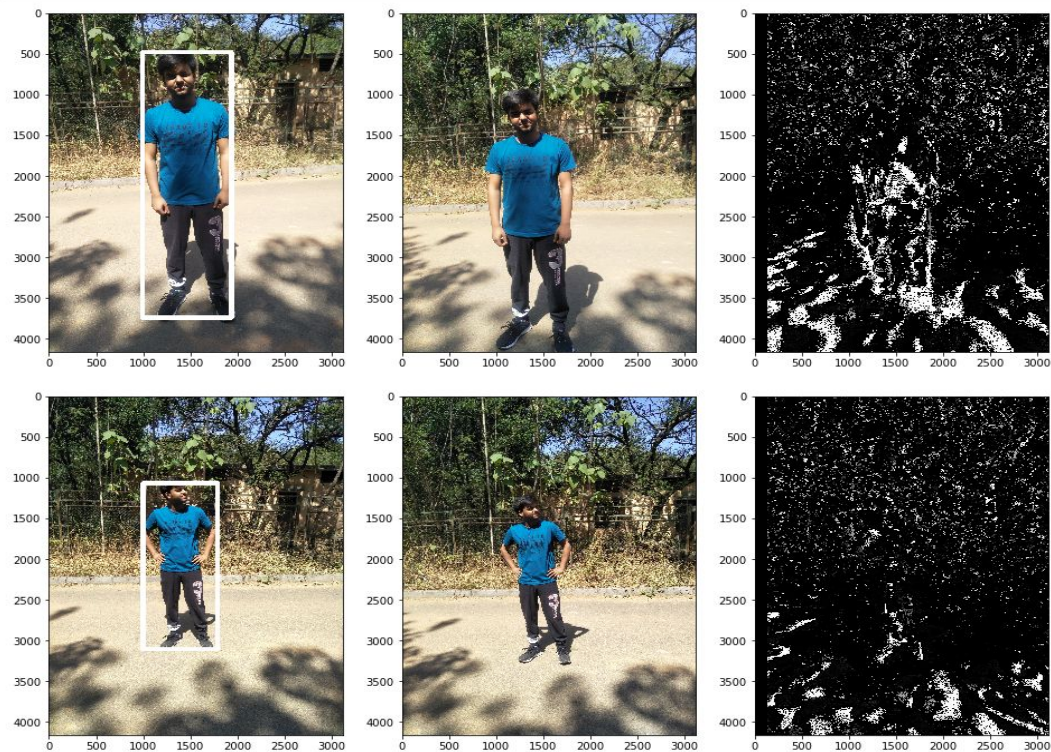
# Experiments



# Experiments

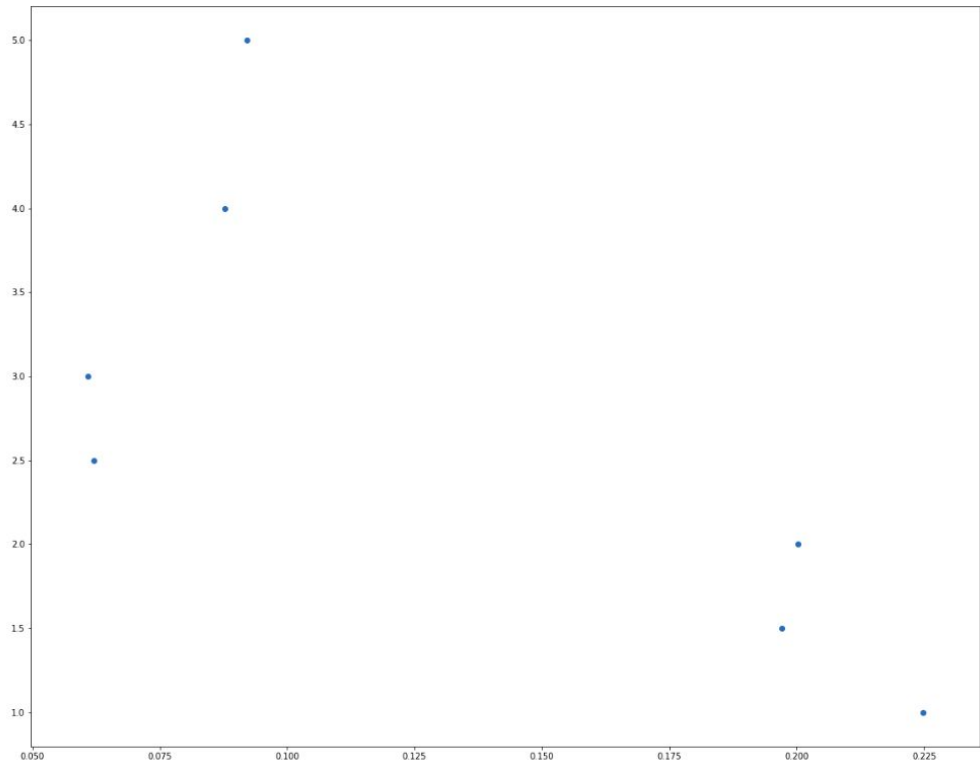


# Experiments

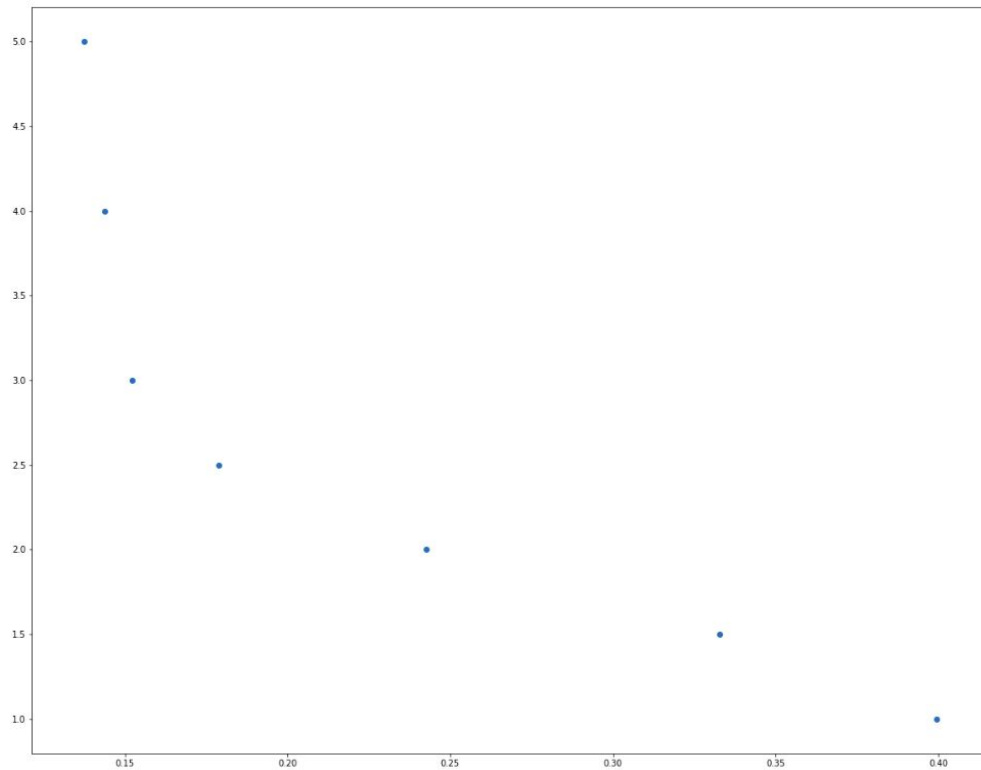




# Experiments



# Experiments



# Experiments

