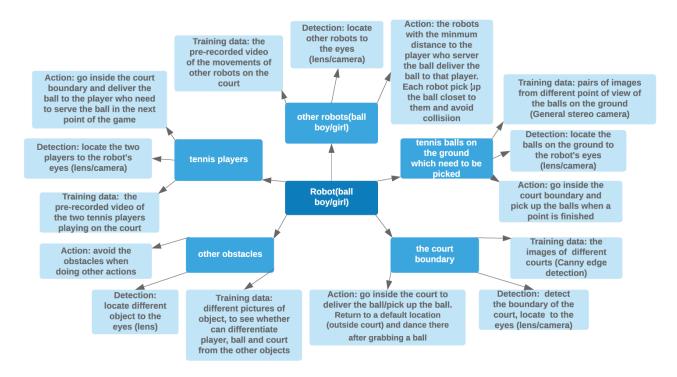
Name: Yuhan Shao UTORid: shaoyuha

Student Number: 1002281327

# CSC420 Assignment4

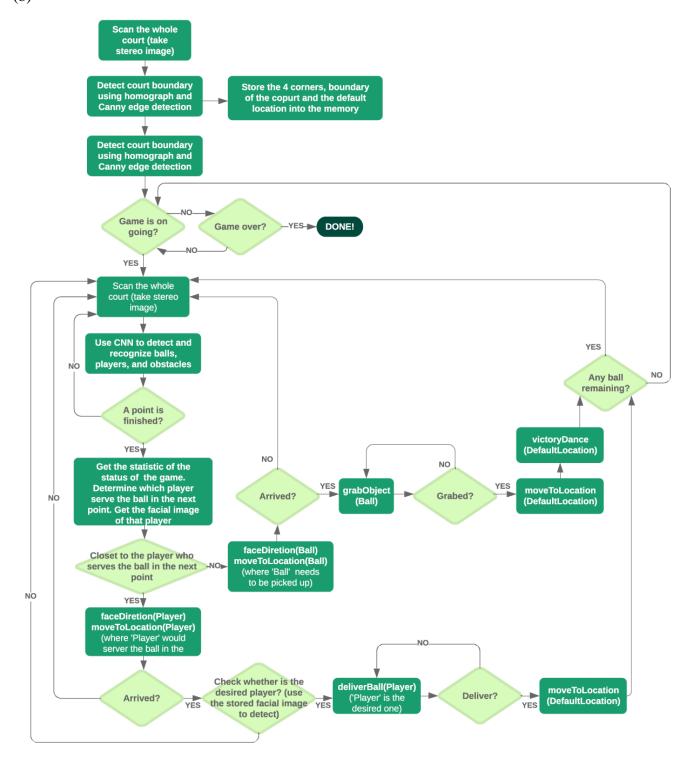
## 1. (a)



# **■** Challenges:

- Other robots: when robots pick up the ball when a point is finished, robots need to cooperate with each other and do not collide with each other.
- Tennis ball on the ground which need to be picked: the robot needs to know the object on the ground is the ball that need to be picked when a point is finished.
- The court boundary: need to use Canny edge detection and homograph to determine the boundary.
- Other obstacles: differentiate other objects with balls, player, court and other robots.
- Tennis players: need to know which player serves the ball in the next point of the game in order to deliver the ball to him.

(b)



# (c) CNN ALGO Idea: https://arxiv.org/pdf/1506.01195.pdf

```
# helper function
CNNTraining(dataSet):
for i in range(size(dataSet)):
    matches = CNN(dataSet.image[i])
    while matches != dataSet.matches:
    backPropagation()
```

```
# consider the default methods: faceDirection(X,Y,Z), moveToLocation(X,Y,Z),
# grabObject(X,Y,Z), victoryDanceAtLocation(X,Y,Z) as the methods of robot class
init robot class
init game class

robot.ballFeatures = CNNTraining(dataSetOfTennisBall)
robot.playersFeatures = CNNTraining(dataSetOfPlayers)
robot.courtBoundary = cannyEdgeDetection(currCourtImage)
```

```
TennisGame(robot, game):
   Pass in the initialized robot and game objects.
   Play the tennis match.
   while !game.gameOver:
       robot.takeStereoImage()
       balls, players, obstacles = robot.recognizeBallsLocation(),
                                 robot.recognizePlayersLocation(),
                                 robot.recognizeObsatclesLocation()
       numRemainingBalls = len(balls)
       if game.pointFinished && numRemainingBalls != 0:
           robot.getCurrentGameStatistic()
           # get the position of the player who serve the ball in the next point
           playerServeBall = robot.getPlayerServeBall()
           # determine whether the current robot is closest to playerServeBall among all other robots
           if robot.closestToPlayer(playerServeBall, robot.otherRobots)
               robot.faceDirection(playerServeBall)
               robot.moveToLocation(playerServeBall)
               # use the facial detection to check
              if robot.arrive(playerServeBall) && robot.checkCorrectPlayer():
                  if robot.deliverBall(playerServeBall):
                      # defaultLocation is somewhere outside of the court, i.e.: the start position and dance position of the robot
                      # here robot not grab the ball, therefore, not dance
                      robot.faceDirection(robot.defaultLocation)
                      robot.moveToLocation(robot.defaultLocation)
           else:
               # the robot grab one ball from the balls on the ground and dance
              ball = robot.ballToGrab(balls)
              robot.faceDirection(ball)
              robot.moveToLocation(ball)
              if robot.arrive(ball):
                  numRemainingBalls--
                  if robot.grabObject(ball):
                      robot.faceDirection(robot.defaultLocation)
                      robot.moveToLocation(robot.defaultLocation)
                      robot.victoryDanceAtLocation(robot.defaultLocation)
```

(d)

- **Need to modify (software):** the general flow chart does not need to change but: Need to change 'robot.takeStereoImage()' method of the robot class.
- Why those changes are important to solve this issue:

  Lose one of its lens, the robot cannot have the two (from 2 lens) stereo images from different point of views simultaneously. Instead, only use one eye (len) to take an image from different point of view sequentially. (i.e.: robot needs to change the face direction in order to change the point of view, and then take an image)
- Conditions that the robot may still be unable to complete its task:

  Problem comes when the object moves (especially when moving fast) during the time when robot changes the face direction and take the images sequentially. The robot would calculate wrong depth/distance to the object from the stereo images it takes. Consequently, robot may crash into the moving objects (e.g.: moving balls, players and obstacles)

## 2. (a) **Code:**

```
import cv2 as cv
import numpy as np

# _ALL CALIB.txt from first 3 images
calib = {'f': 721.537700, 'px': 609.559300, 'py': 172.854000, 'baseline':
0.5327119288}

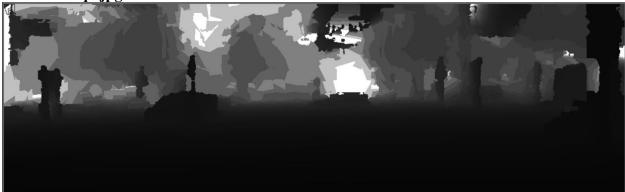
def calculate_depth(imgs, calib):
    disparity = cv.imread(imgs)
    depth = np.true_divide(calib['f'] * calib['baseline'], disparity)
    return depth

if __name__ == '__main__':
    cv.imwrite('004945q1.jpg', calculate_depth('004945_left_disparity.png', calib))
    cv.imwrite('004964q1.jpg', calculate_depth('004964_left_disparity.png', calib))
    cv.imwrite('005002q1.jpg', calculate_depth('005002_left_disparity.png', calib))
```

- '004945q1.jpg':



'004964q1.jpg':



- '005002q1.jpg':



(b)

- Reuse the output 'output\_dict' from the function 'run\_inference\_for\_single\_image(image, graph)'
- Eliminate the data from the 'output\_dict' according to the 'threshold' variable. Adjust the 'threshold' variable to order to keep approximate 3-5 'num\_detections' ('threshold' = 0.5)

- Use the following function to eliminate the dictionary

```
import numpy as np
threshold = 0.5
def eliminate dict(output dict):
    scores = output dict['detection scores']
    index = []
    num detection = 0
    for i in range(output_dict['num_detections']):
         score = scores[i]
         if score > threshold:
              index.append(i)
              num detection += 1
    detection_boxes = []
    detection_scores = []
    detection_classes = []
eliminated_dict = {}
     for i in range(len(index)):
         detection boxes.append(output dict['detection boxes'][index[i]])
         detection_scores.append(output_dict['detection_scores'][index[i]])
detection_classes.append(output_dict['detection_classes'][index[i]])
```

```
eliminated_dict['num_detection'] = num_detection
eliminated_dict['detection_boxes'] = np.array(detection_boxes)
eliminated_dict['detection_scores'] = np.array(detection_scores)
eliminated_dict['detection_classes'] = np.array(detection_classes)

return eliminated_dict
```

- Then, modify the main function in 'object\_detection\_tutorial.ipynb' as below:

```
for image path in TEST IMAGE PATHS:
  image = Image.open(image_path)
  # the array based representation of the image will be used later in order to prepare the
  # result image with boxes and labels on it.
  image_np = load_image_into_numpy_array(image)
  # Expand dimensions since the model expects images to have shape: [1, None, None, 3]
  image_np_expanded = np.expand_dims(image_np, axis=0)
  # Actual detection.
  output_dict = run_inference_for_single_image(image_np, detection_graph)
  # Add the elimination dictionary code
  output dict = eliminate dict(output dict)
  print(output dict)
  # Visualization of the results of a detection.
  vis_util.visualize_boxes_and_labels_on_image_array(
      output_dict['detection_boxes'],
     output_dict['detection_classes'],
      output_dict['detection_scores'],
     category_index,
      instance masks=output dict.get('detection_masks'),
      use_normalized_coordinates=True,
      line thickness=8)
  plt.figure(figsize=IMAGE_SIZE)
  plt.imshow(image np)
```

- The printed eliminated 'output\_dict' output would as below:

```
{'num detection': 6, 'detection boxes': array([[0.50004214, 0.6555561 , 0.8872184 , 0.8707886
       [0.44908354, 0.584361 , 0.5227398 , 0.6364023 ],
       [0.01041733, 0.5026116 , 0.09855365, 0.5190967 ],
       [0.46349224, 0.42532513, 0.5317025 , 0.47155157],
       [0.4670957, 0.35683563, 0.53998923, 0.3719673],
       [0.49316233, 0.23185101, 0.54320043, 0.26347476]], dtype=float32), 'detection scores':
array([0.9491954 , 0.903302 , 0.7825848 , 0.72892267, 0.66242313,
       0.57823133], dtype=float32), 'detection classes': array([ 3, 3, 10, 3, 1, 3], dtyp
e=uint8) }
{'num detection': 6, 'detection boxes': array([[0.47048816, 0.44235417, 0.5088938 , 0.474377
       [0.46128154, 0.5218015 , 0.55010664, 0.58383596],
       [0.4213005, 0.7209805, 0.5365165, 0.80251265],
       [0.47370738, 0.49366155, 0.520934, 0.5202959],
       \hbox{\tt [0.5039439 , 0.15278855, 0.5627457 , 0.21511334],}
       [0.28877726, 0.29553753, 0.3853735 , 0.3158021 ]], dtype=float32), 'detection scores':
array([0.88989836, 0.7733079 , 0.7360068 , 0.72747695, 0.65513057,
       0.555762 ], dtype=float32), 'detection classes': array([ 3,
                                                                       3, 3, 3, 10], dtyp
{'num detection': 4, 'detection boxes': array([[0.46915835, 0.39268324, 0.59815925, 0.5128745
       [0.49698436, 0.16895944, 0.70170236, 0.31567362],
       [0.41786578, 0.8332899 , 0.6043134 , 0.986494
[0.0924626, 0.814217, 0.18619645, 0.8461287]], dtype=float32), 'detection scores': array([0.88686144, 0.8019987, 0.72281444, 0.50532067], dtype=float32), 'detection_classes':
array([ 3, 3, 10], dtype=uint8)}
```

- (c)
- Reuse the eliminated dictionary from the previous question.
- → Store the dictionary data in the text file as the following format:

■ datafile\_dict1.txt

```
0.50004214, 0.6555561 , 0.8872184 , 0.8707886
0.46349224\,,\; 0.42532513\,,\; 0.5317025\,\;,\; 0.47155157
 0.4670957 \ , \ 0.35683563 , \ 0.53998923 , \ 0.3719673 
0.49316233, 0.23185101, 0.54320043, 0.26347476
0.9491954
0.903302
0.7825848
0.72892267
0.66242313
0.57823133
3
3
10
3
1
3
```

■ datafile\_dict2.txt

```
6
0.47048816, 0.44235417, 0.5088938, 0.474377
0.46128154, 0.5218015, 0.55010664, 0.58383596
0.4213005, 0.7209805, 0.5365165, 0.80251265
0.47370738, 0.49366155, 0.520934, 0.5202959
0.5039439, 0.15278855, 0.5627457, 0.21511334
0.28877726, 0.29553753, 0.3853735, 0.3158021
0.88989836
0.7733079
0.7360068
0.72747695
0.65513057
0.5555762
3
3
3
3
3
3
3
3
10
```

## ■ datafile\_dict3.txt

```
4
0.46915835, 0.39268324, 0.59815925, 0.51287454
0.49698436, 0.16895944, 0.70170236, 0.31567362
0.41786578, 0.8332899, 0.6043134, 0.986494
0.0924626, 0.814217, 0.18619645, 0.8461287
0.88686144
0.8019987
0.72281444
0.50532067
3
3
3
10
```

→ Then, read the data from the text file and store it as dictionary:

```
read data(filename):
img_dict = {}
file = open(filename, 'r')
num detection = int(file.readline())
imm_detection = Int(!TetricateInt())
img_dict['num_detection'] = num_detection
img_dict['detection_boxes'] = []
img_dict['detection_scores'] = []
img dict['detection classes'] = []
# read box
for i in range(num detection):
     line box = file.readline()
     box = line box.split (',')
     for j in range(4):
     box[j] = float(box[j])
img_dict['detection_boxes'].append(box)
# read score
for i in range(num detection):
     line score = file.readline()
     img dict['detection scores'].append(float(line score))
# read type
for i in range(num detection):
     line type = file.readline()
     img_dict['detection_classes'].append(int(line_type))
return img_dict
_name__ == '__main__':
img1_dict = read_data('datafile_dict1.txt')
print(img1 dict)
img2 dict = read data('datafile dict2.txt')
print(img2 dict)
img3_dict = read_data('datafile_dict3.txt')
print(img3_dict)
```

## ■ Output:

 $\{ \text{'num\_detection': 6, 'detection\_boxes': } [[0.50004214, 0.6555561, 0.8872184, 0.8707886], [0.44908354, 0.584361, 0.5227398, 0.6364023], [0.01041733, 0.5026116, 0.09855365, 0.5190967], [0.46349224, 0.42532513, 0.5317025, 0.47155157], [0.4670957, 0.35683563, 0.53998923, 0.3719673], [0.49316233, 0.23185101, 0.54320043, 0.26347476]], 'detection\_scores': [0.9491954, 0.903302, 0.7825848, 0.72892267, 0.66242313, 0.57823133], 'detection\_classes': [3, 3, 10, 3, 1, 3] \}$ 

 $\begin{tabular}{ll} & \{ \text{'num\_detection': 6, 'detection\_boxes': } [[0.47048816, 0.44235417, 0.5088938, 0.474377], [0.46128154, 0.5218015, 0.55010664, 0.58383596], [0.4213005, 0.7209805, 0.5365165, 0.80251265], [0.47370738, 0.49366155, 0.520934, 0.5202959], [0.5039439, 0.15278855, 0.5627457, 0.21511334], [0.28877726, 0.29553753, 0.3853735, 0.3158021]], 'detection\_scores': [0.88989836, 0.7733079, 0.7360068, 0.72747695, 0.65513057, 0.555762], 'detection\_classes': [3, 3, 3, 3, 3, 10] \}$ 

 $\begin{tabular}{ll} & \{ num\_detection': 4, 'detection\_boxes': [[0.46915835, 0.39268324, 0.59815925, 0.51287454], \\ [0.49698436, 0.16895944, 0.70170236, 0.31567362], [0.41786578, 0.8332899, 0.6043134, 0.986494], \\ [0.0924626, 0.814217, 0.18619645, 0.8461287]], 'detection\_scores': [0.88686144, 0.8019987, 0.72281444, 0.50532067], 'detection\_classes': [3, 3, 3, 10] \} \end{tabular}$ 

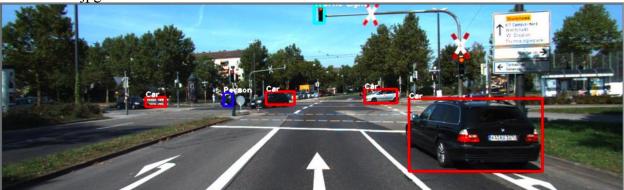
 $\rightarrow$  In the following question's code, consider the dictionary that we constructed from the data in the text file as the global variable. (i.e.:  $img1\_dict$ ,  $img2\_dict$ ,  $img3\_dict$ )

→ Now, consider the following code:

```
import cv2 as cv
[0.46349224, 0.42532513, 0.5317025 , 0.47155157],
         [0.4670957, 0.35683563, 0.53998923, 0.3719673], [0.49316233, 0.23185101, 0.54320043, 0.26347476]]
'detection_scores': [0.9491954 , 0.903302 , 0.7825848 , 0.72892267, 0.66242313, 0.57823133],
    'detection_classes': [3, 3, 10, 3, 1, 3]}
img2_dict = {'num_detection': 6, 'detection_boxes': [[0.47048816, 0.44235417, 0.5088938 , 0.474377 ],
    [0.46128154, 0.5218015 , 0.55010664, 0.58383596],
    [0.4213005 , 0.7208805 , 0.5365165 , 0.80251265]
          0.4213005 , 0.7209805 , 0.5365165 , 0.80251265],
                                                     , 0.5202959
          0.47370738, 0.49366155, 0.520934
         [0.5639439 , 0.15278855 , 0.5627457 , 0.21511334],
[0.28877726 , 0.29553753 , 0.3853735 , 0.3158021 ]],
'detection_scores': [0.88989836 , 0.7733079 , 0.7360068 , 0.72747695 , 0.65513057 , 0.555762 ],
'detection_classes': [ 3, 3, 3, 3, 10]}
img3_dict = {'num_detection': 4, 'detection_boxes': [[0.46915835, 0.39268324, 0.59815925, 0.51287454],
          [0.4969843\overline{6}, 0.16895944, 0.70170236, 0.31567362],
          0.41786578, 0.8332899 , 0.6043134 , 0.986494 ],
[0.0924626 , 0.814217 , 0.18619645, 0.8461287 ]]
         [0.0924626 , 0.814217
           'detection_scores': [0.88686144, 0.8019987 , 0.72281444, 0.50532067],
          'detection_classes': [ 3, 3, 3, 10]}
def visualize(img, img_dict):
     img = cv.imread(img)
     height, width = img.shape[0], img.shape[1]
     for i in range(img_dict['num_detection']):
    box = img_dict['detection_boxes'][i]
    pt1 = (int(box[1] * width), int(box[0] * height))
    pt2 = (int(box[3] * width), int(box[2] * height))
          classes = img_dict['detection_classes'][i]
          if classes == 1:
               # person, blue (b, g, r)
               cv.rectangle(img, pt1, pt2, color=(255, 0, 0), thickness=3)
               cv.putText(img, 'Person', org=pt1, color=(255, 255, 255), fontScale=1, fontFace=1,
thickness=2)
          elif classes == 2:
               # bicycle, green
               cv.rectangle(img, pt1, pt2, color=(0, 255, 0), thickness=3)
               cv.putText ( img, 'Bicycle', org=pt1, color=(255, 255, 255), fontScale=1, fontFace=1,
thickness=2
          elif classes == 3:
               # car, red
               cv.rectangle(img, pt1, pt2, color=(0, 0, 255), thickness=3)
               cv.putText(img, 'Car', org=pt1, color=(255, 255, 255), fontScale=1, fontFace=1,
thickness=2)
          elif classes == 10:
               # traffic light, cyan
               cv.rectangle(img, pt1, pt2, color=(255, 255, 0), thickness=3)
cv.putText(img, 'Traffic Light', org=pt1, color=(255, 255, 255), fontScale=1, fontFace=1,
thickness=2)
     return ima
               == '
if __name_
                      main
     cv.imwrite('005002q2_3.jpg', visualize('005002.jpg', img3_dict))
```

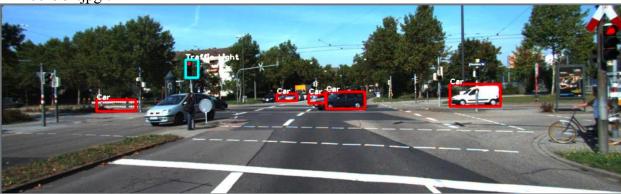
# - Output of the 3 images:

■ '004945.jpg':

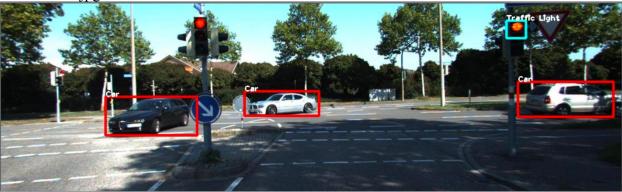


(Note: 'traffic light' text has been written on the top, but slightly out of image boundary)

• '004964.jpg':



**■** '005002.jpg':



(d)

Compute the 3D location (center of mass) of each detected object:

- For the first 3 images, we have the dictionary of data for the objects on the image to be detected (i.e.: person, bicycle, car, traffic light).
- For each pixel inside the boundary box, first, calculate the (x, y, z) of each pixel in the real world, store as (z, x, y) in a list, then sort the list according to the z (depth) value. Then, get the **MEAN depth** (i.e.: the middle depth/z value of the sorted list of tuple) among all pixel, which is considered as the center of mass.

### - Code:

```
import cv2 as cv
import numpy as np
imgl dict = {'num detection': 6, 'detection boxes': [[0.50004214, 0.6555561 , 0.8872184 ,
0.8707886 1.
          [0.44908354, 0.584361 , 0.5227398 , 0.6364023 ], [0.01041733, 0.5026116 , 0.09855365, 0.5190967 ], [0.46349224, 0.42532513, 0.5317025 , 0.47155157],
          [0.4670957 , 0.35683563, 0.53998923, 0.3719673 ], [0.49316233, 0.23185101, 0.54320043, 0.26347476]],
            'detection scores': [0.9491954 , 0.903302 , 0.7825848 , 0.72892267, 0.66242313,
0.57823133]
            'detection_classes': [ 3, 3, 10, 3, 1, 3]}
img2 dict = {'num detection': 6, 'detection boxes': [[0.47048816, 0.44235417, 0.5088938 ,
           [0.46128154, 0.5218015, 0.55010664, 0.58383596],
           [0.4213005, 0.7209805, 0.5365165, 0.80251265],
          [0.47370738, 0.49366155, 0.520934 , 0.5202959 ], [0.5039439 , 0.15278855, 0.5627457 , 0.21511334], [0.28877726, 0.29553753, 0.3853735 , 0.3158021 ]]
            'detection scores': [0.88989836, 0.7733079 , 0.7360068 , 0.72747695, 0.65513057,
              ],
'detection_classes': [ 3, 3, 3, 10]}
img3_dict = {'num_detection': 4, 'detection_boxes': [[0.46915835, 0.39268324, 0.59815925,
0.51\overline{2}87454].
           [0.49698436, 0.16895944, 0.70170236, 0.31567362],
          [0.41786578, 0.8332899, 0.6043134, 0.986494], [0.0924626, 0.814217, 0.18619645, 0.8461287]],
'detection_scores': [0.88686144, 0.8019987 , 0.72281444, 0.50532067],
    'detection_classes': [ 3,  3,  3,  10]}
calib = {'f': 721.537700, 'px': 609.559300, 'py': 172.854000, 'baseline': 0.5327119288}
def calculate depth(img):
     disparity = cv.imread(img)
      depth = np.true divide(calib['f'] * calib['baseline'], disparity)
      return depth
def center(ZDepth, img, img_dict):
      img = cv.imread(img)
      height, width = img.shape[0], img.shape[1]
      box depth = []
      for i in range(img dict['num detection']):
            depth = []
           box = img_dict['detection_boxes'][i]
pt1 = (int(box[1] * width), int(box[0] * height))
pt2 = (int(box[3] * width), int(box[2] * height))
            for row in range(pt1[1], pt2[1]):
                  for column in range(pt1[0], pt2[0]):
                       z = ZDepth[row, column, 0]
                       x = np.true_divide((column - calib['px']) * z, calib['f'])
                       y = np.true_divide((row - calib['py']) * z, calib['f'])
                       depth.append((z, x, y))
            depth.sort()
            box depth.append(depth[len(depth) // 2]) # (z, x, y)
      print(box depth)
      return box_depth
if __name__ == '__main__':
    ZDepth1 = calculate_depth('004945_left_disparity.png')
    ZDepth2 = calculate_depth('004964_left_disparity.png')
    ZDepth3 = calculate_depth('005002_left_disparity.png')
     center(ZDepth1, '004945_left_disparity.png', img1_dict)
center(ZDepth2, '004964_left_disparity.png', img2_dict)
center(ZDepth3, '005002_left_disparity.png', img3_dict)
```

# - Output: (the 'box\_depth' of the 3 images, list of (z, x, y) tuple)

```
[(7.252296978658787, 3.381622155166456, 0.4135653777812225), (48.046467483614464, 11.6158302197777, 1.075145850300599), (20.230091572048195, 0.8815176810432712, -4.0893771401681684), (42.70797109654619, -3.4069473025975374, 0.4229732714671992), (42.70797109654619, -9.089207876464204, 1.1924460575116436), (76.87434797378314, -33.51389828499557, 1.5071485889609586)]

[(76.87434797378314, -4.214742200995568, 0.4417247313609585), (34.9428854426287, 3.3144706641292867, 0.3460690402913448), (38.43717398689157, 18.934750087102216, 0.48721833008047927), (64.06195664481929, 1.1045515487703592, 0.7232452286674654), (54.91024855270224, -26.449855010768257, 2.522467084572113), (16.711814776909378, -5.293750676007732, -1.3168175652171827)]

[(25.624782657927714, -0.8722021381985228, 0.5734111201603195), (16.015489161204822, -6.338368562074076, 1.2462351647668664), (19.218586993445786, 13.888884052591107, 0.003888797080239642), (8.355907388454689, 4.950059994448307, -1.2606048760346782)]
```

### (e)

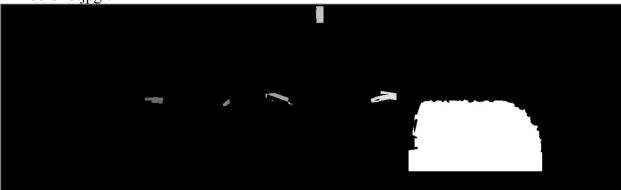
### - Code:

```
import cv2 as cv
import numpy as np
imgl dict = {'num detection': 6, 'detection boxes': [[0.50004214, 0.6555561 , 0.8872184 ,
0.8707886 1
                  [0.44908354, 0.584361, 0.5227398, 0.6364023],
                   [0.01041733, 0.5026116 , 0.09855365, 0.5190967 ],
                   [0.46349224, 0.42532513, 0.5317025 , 0.47155157],
                  [0.4670957 , 0.35683563, 0.53998923, 0.3719673 ]
                  [0.49316233, 0.23185101, 0.54320043, 0.26347476]]
                     'detection scores': [0.9491954 , 0.903302 , 0.7825848 , 0.72892267, 0.66242313,
0.578231331.
'detection_classes': [ 3, 3, 10, 3, 1, 3]}  img2\_dict = \{ 'num\_detection': 6, 'detection\_boxes': [[0.47048816, 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.5088938 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.44235417, 0.508898 , 0.4488818 , 0.448888 , 0.448888 , 0.448888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.44888 , 0.448
0.47\overline{4377}
                  [0.46128154, 0.5218015, 0.55010664, 0.58383596],
                   [0.4213005 , 0.7209805 , 0.5365165 , 0.80251265],
                  [0.47370738, 0.49366155, 0.520934 , 0.5202959 ],
                  [0.5039439 , 0.15278855, 0.5627457 , 0.21511334],
                  [0.28877726, 0.29553753, 0.3853735 , 0.3158021 ]]
                     'detection scores': [0.88989836, 0.7733079 , 0.7360068 , 0.72747695, 0.65513057,
'detection_classes': [ 3, 3, 3, 3, 10]} img3_dict = {'num_detection': 4, 'detection_boxes': [[0.46915835, 0.39268324, 0.59815925,
0.51287454],
                   [0.49698436, 0.16895944, 0.70170236, 0.31567362],
                   [0.41786578, 0.8332899 , 0.6043134 , 0.986494
                  [0.0924626 , 0.814217 , 0.18619645 , 0.8461287 ]],
'detection_scores': [0.88686144, 0.8019987, 0.72281444, 0.50532067],
    'detection_classes': [3, 3, 3, 10]}
calib = {'f': 721.537700, 'px': 609.559300, 'py': 172.854000, 'baseline': 0.5327119288}
def calculate_depth(img):
         disparity = cv.imread(img)
          depth = np.true divide(calib['f'] * calib['baseline'], disparity)
          return depth
def center(ZDepth, img, img dict):
          img = cv.imread(img)
```

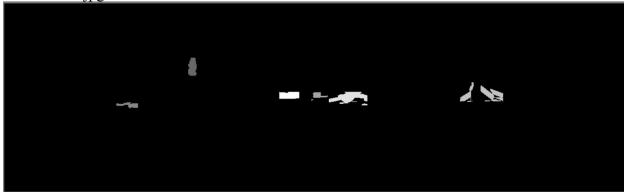
```
height, width = img.shape[0], img.shape[1]
    box depth = []
     for i in range(img_dict['num_detection']):
         depth = []
         box = img dict['detection boxes'][i]
         pt1 = (int(box[1] * width), int(box[0] * height))
pt2 = (int(box[3] * width), int(box[2] * height))
         for row in range(pt1[1], pt2[1]):
              for column in range(pt1[0], pt2[0]):
                   z = ZDepth[row, column, 0]
                   x = np.true divide((column - calib['px']) * z, calib['f'])
                   y = np.true_divide((row - calib['py']) * z, calib['f'])
                   depth.append((z, x, y))
         depth.sort()
         box depth.append(depth[len(depth) // 2]) # (z, x, y)
    print(box depth)
     return box depth
def segmentation(img, ZDepth, box_depth, img_dict):
     img = cv.imread(img)
    height, width = img.shape[0], img.shape[1]
    segmentation = np.zeros((height, width, 3))
    num_detection = img_dict['num_detection']
     for i in range(num_detection):
         box = img_dict['detection_boxes'][i]
         pt1 = (int(box[1] * width), int(box[0] * height))
pt2 = (int(box[3] * width), int(box[2] * height))
         for row in range(pt1[1], pt2[1]):
              for column in range(pt1[0], pt2[0]):
                   z = ZDepth[row, column, 0]
                   x = np.true_divide((column - calib['px']) * z, calib['f'])
y = np.true_divide((row - calib['py']) * z, calib['f'])
                   box_depth[i][2]) ** 2 <= 9:
                        segmentation[row, column] = 255 - 30 * i
     return segmentation
if __name__ == '__main__':
     ZDepth1 = calculate depth('004945 left disparity.png')
    ZDepth2 = calculate depth('004964 left disparity.png')
    ZDepth3 = calculate_depth('005002_left_disparity.png')
    box_depth1 = center(ZDepth1, '004945_left_disparity.png', img1_dict)
box_depth2 = center(ZDepth2, '004964_left_disparity.png', img2_dict)
box_depth3 = center(ZDepth3, '005002_left_disparity.png', img3_dict)
    cv.imwrite('004945seg.jpg', segmentation('004945.jpg', ZDepth1, box_depth1, img1_dict))
    cv.imwrite('004964seg.jpg', segmentation('004964.jpg', ZDepth2, box_depth2, img2_dict))
cv.imwrite('005002seg.jpg', segmentation('005002.jpg', ZDepth3, box_depth3, img3_dict))
```

# - Output:

• '004945.jpg':



• '004964.jpg':



■ '005002.jpg':



(f)

# **■** Code:

```
import cv2 as cv
import numpy as np

img1_dict = {'num_detection': 6, 'detection_boxes': [[0.50004214, 0.6555561 , 0.8872184 ,
0.8707886 ],
       [0.44908354, 0.584361 , 0.5227398 , 0.6364023 ],
       [0.01041733, 0.5026116 , 0.09855365, 0.5190967 ],
       [0.46349224, 0.42532513, 0.5317025 , 0.47155157],
       [0.4670957 , 0.35683563, 0.53998923, 0.3719673 ],
       [0.49316233, 0.23185101, 0.54320043, 0.26347476]],
       'detection_scores': [0.9491954 , 0.903302 , 0.7825848 , 0.72892267, 0.66242313,
0.57823133],
```

```
'detection_classes': [ 3, 3, 10, 3, 1, 3]}
img2 dict = {'num detection': 6, 'detection boxes': [[0.47048816, 0.44235417, 0.5088938 ,
0.47\overline{4}377
        [0.46128154, 0.5218015, 0.55010664, 0.58383596],
        [0.4213005, 0.7209805, 0.5365165, 0.80251265],
        [0.47370738, 0.49366155, 0.520934 , 0.5202959 ], [0.5039439 , 0.15278855, 0.5627457 , 0.21511334], [0.28877726, 0.29553753, 0.3853735 , 0.3158021 ]],
         'detection scores': [0.88989836, 0.7733079 , 0.7360068 , 0.72747695, 0.65513057,
0.555762
'detection_classes': [ 3, 3, 3, 10]}
img3_dict = {'num_detection': 4, 'detection_boxes': [[0.46915835, 0.39268324, 0.59815925,
0.\overline{5}1\overline{2}87454],
        [0.49698436, 0.16895944, 0.70170236, 0.31567362],
        [0.41786578, 0.8332899 , 0.6043134 , 0.986494 ], [0.0924626 , 0.814217 , 0.18619645, 0.8461287 ]],
         'detection_scores': [0.88686144, 0.8019987 , 0.72281444, 0.50532067], 'detection_classes': [ 3,  3,  3,  10]}
calib = {'f': 721.537700, 'px': 609.559300, 'py': 172.854000, 'baseline': 0.5327119288}
def num object(img dict):
     count_person, count_bicycle, count_car, count_traffic_light = 0, 0, 0, 0
    type_dict = {}
     for i in range(img dict['num detection']):
         classes = img dict['detection classes'][i]
         if classes == 1:
              # person
              count person += 1
              if 1 in type_dict:
                   type_dict[1].append(i)
              else:
                  type dict[1] = [i]
         elif classes == 2:
              # bicycle
              count bicycle += 1
              if 2 in type dict:
                   type_dict[2].append(i)
              else:
                   type dict[2] = [i]
         elif classes == 3:
              # car
              count_car += 1
if 3 in type_dict:
                  type_dict[3].append(i)
                  type_dict[3] = [i]
         elif classes == 10:
              # traffic_light
              count_traffic_light += 1
              if 10 in type_dict:
                   type dict[4].append(i)
              else:
                   type_dict[4] = [i]
    strl = 'There is(are) {} person in the scene; {} bicycle(s) in the scene; {} car(s) in the
scene.'.format(
         count person, count bicycle, count car)
    print(str1)
    if count traffic light != 0:
         str2 = 'There is(are) {} traffic light nearby.'.format(count traffic light)
         print(str2)
     return type dict
```

```
def calculate_depth(img): # input is left_disparity
    disparity = cv.imread(img)
    depth = np.true divide(calib['f'] * calib['baseline'], disparity)
    return depth
def center(ZDepth, img, img_dict):
    Return list of (z, x, y) of the center point of the object inside the box.
    img = cv.imread(img)
    height, width = img.shape[0], img.shape[1]
    box_depth = []
    for i in range(img dict['num detection']):
         depth = []
         box = img_dict['detection_boxes'][i]
        pt1 = (int(box[1] * width), int(box[0] * height))
pt2 = (int(box[3] * width), int(box[2] * height))
         for row in range(pt1[1], pt2[1]):
             for column in range(pt1[0], pt2[0]):
                 z = ZDepth[row, column, 0]
                 x = np.true_divide((column - calib['px']) * z, calib['f'])
y = np.true_divide((row - calib['py']) * z, calib['f'])
                 depth.append((z, x, y))
         depth.sort()
         box depth.append(depth[len(depth) // 2]) # (z, x, y)
    # print(box depth)
    return box depth
def find_closest(img_dict, box_depth, type_dict):
    Find the closest object among all. box depth is list of (z, x, y)
    classes = type_dict.keys()
    for types in classes:
         types_index = type_dict[types] # list of index
         min_index_distance = (0, np.inf)
         for i in types_index:
             pt = box_depth[i]
             distance = (pt[0] ** 2 + pt[1] ** 2 + pt[2] ** 2) ** (1 / 2)
if distance < min_index_distance[1]:</pre>
                 min_index_distance = (i, distance)
         X = box depth[min index distance[0]][1]
         print_helper(min_index_distance[0], X, min_index_distance[1], img_dict)
def print helper(index, X, min distance, img dict):
    if X >= 0:
         txt = 'to your right'
    else:
         txt = 'to your left'
    types = img_dict['detection_classes'][index]
    if types == 1:
         # person
         label = 'person'
    elif types == 2:
         # bicycle
         label = 'bicycle'
    elif types == 3:
         # car
         label = 'car'
    elif types == 10:
         # traffic light
```

```
label = 'traffic light'
   str = 'There is a {} {} meters {}.\n It is {} meters away from you.\n'.format(label, abs(X),
txt, min_distance)
   print(str)
if __name__ == '__main__':
    # image1
   print('======= image1 =======')
   ZDepth1 = calculate_depth('004945 left disparity.png')
   box depth1 = center(ZDepth1, '004945 left disparity.png', img1 dict)
   type_dict1 = num_object(img1_dict)
    find closest(img1 dict, box depth1, type dict1)
   # image2
   print('======= image2 =======')
    .
ZDepth2 = calculate_depth('004964_left_disparity.png')
   box depth2 = center(ZDepth2, '004964 left disparity.png', img2 dict)
    type dict2 = num object(img2 dict)
    find_closest(img2_dict, box_depth2, type_dict2)
   # image3
   print('======= image3 =======')
    .
ZDepth3 = calculate_depth('005002_left_disparity.png')
   box depth3 = center(ZDepth3, '005002 left disparity.png', img3 dict)
    type dict3 = num object(img3 dict)
    find closest(img3 dict, box depth3, type dict3)
```

```
■ Output:
               = image1 ==:
There is(are) 1 person in the scene; 0 bicycle(s) in the scene; 4 car(s) in the scene.
There is(are) 1 traffic light nearby.
There is a car 3.381622155166456 meters to your right.
It is 8.012628544284036 meters away from you
There is a traffic light 0.8815176810432712 meters to your right.
It is 20.658090033446395 meters away from you
There is a person 9.089207876464204 meters to your left.
It is 43.68073285334254 meters away from you
======= image2 ========
There is(are) 0 person in the scene; 0 bicycle(s) in the scene; 5 car(s) in the scene.
There is(are) 1 traffic light nearby.
There is a car 3.3144706641292867 meters to your right.
It is 35.10143476584247 meters away from you
There is a traffic light 5.293750676007732 meters to your left.
It is 17.579606305532653 meters away from you
====== image3 ======
There is(are) 0 person in the scene; 0 bicycle(s) in the scene; 3 car(s) in the scene.
There is(are) 1 traffic light nearby.
There is a car 6.338368562074076 meters to your left.
It is 17.26916069724466 meters away from you
There is a traffic light 4.950059994448307 meters to your right.
It is 9.793539037883681 meters away from you
```