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MECH6631 Project Report

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Design and implement an Image Processing algorithm to locate and identify all objects in the map. Design and implement an Hiding Strategy to find the closest safe point.

Xiaobo Wu(40216033)

Design and implement a path planning algorithm to find path point from current position to expected position and design a hiding algorithm with Qiaomeng Qin and write the report of these two parts.

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Design and implement a controller to move the robot to any desired location. Also, implement a strategy for the attack task.

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Challenge Level Assumptions to be Removed

1. obstacles of known size. 2. obstacles of circular shape. 3. stationary obstacles.

All files for this project is accessible at:

https://github.com/AreteQin/MECH_Project

2022.04.01

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Chapter 1

Introduction

As shown in Figure 1.1, two robots perform a competition in this project. Two robots chase each other and try to hit the opponent with laser, which are controlled via intelligent algorithms. This report mainly introduces the algorithms for image processing and robot control.

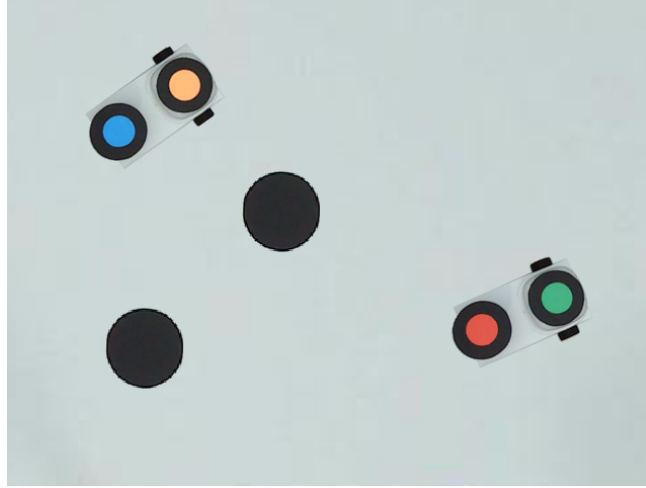


Figure 1.1: Overview of the project.

We assume that there is no wheel slipping,

$$v_r = \omega_r R$$

Where v_r is the linear velocity, R is the radius of wheel, and ω_r is the angular velocity.

x_c and y_c is the coordinate of the vehicle centre. θ is the direction of vehicle. D is the distance between two wheels. The geometry model of this vehicle is shown below:

$$v_c = (v_r + v_l)/2$$

$$\omega_c = (v_r - v_l)/D$$

$$\dot{\theta}_c = \omega_c = (v_r - v_l)/D$$

Chapter 2

Image Processing

2.1 Locating Objects

The flowchart of whole image processing algorithm is shown in figure 2.1.

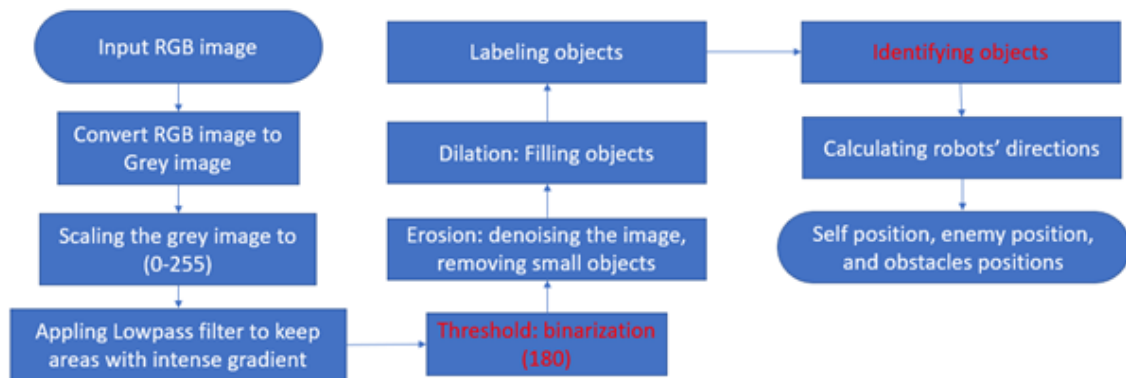


Figure 2.1: Image processing flowchart.

While testing the code, we found if color of obstacle is changed to blue, the threshold of binarization, with original value of 80, would be not suitable anymore. After testing, the threshold is determined as 180, which is suitable for all colors of obstacles. The comparison of different thresholds is shown in figure 2.2.

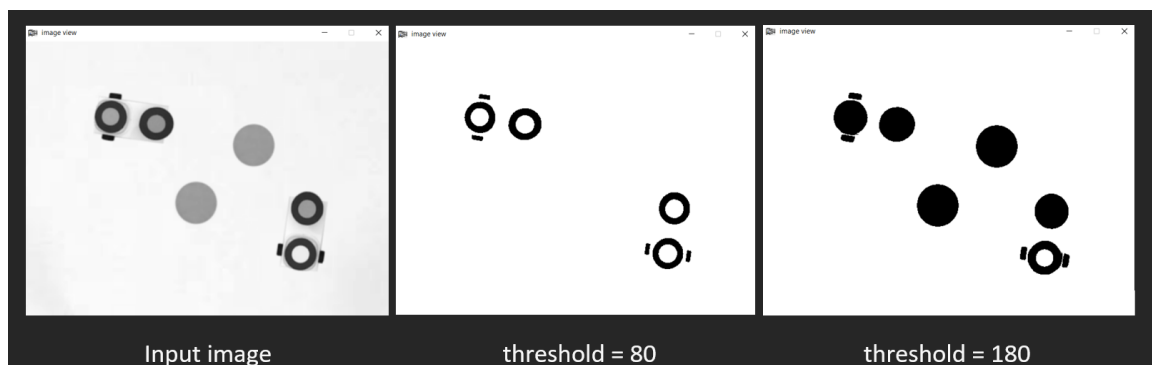


Figure 2.2: Different thresholds in binarization.

2.2 Identifying Objects

Identifying objects is the key of it, the flowchart is shown in figure 2.3 (assume that self-robot is robot A, red and green)

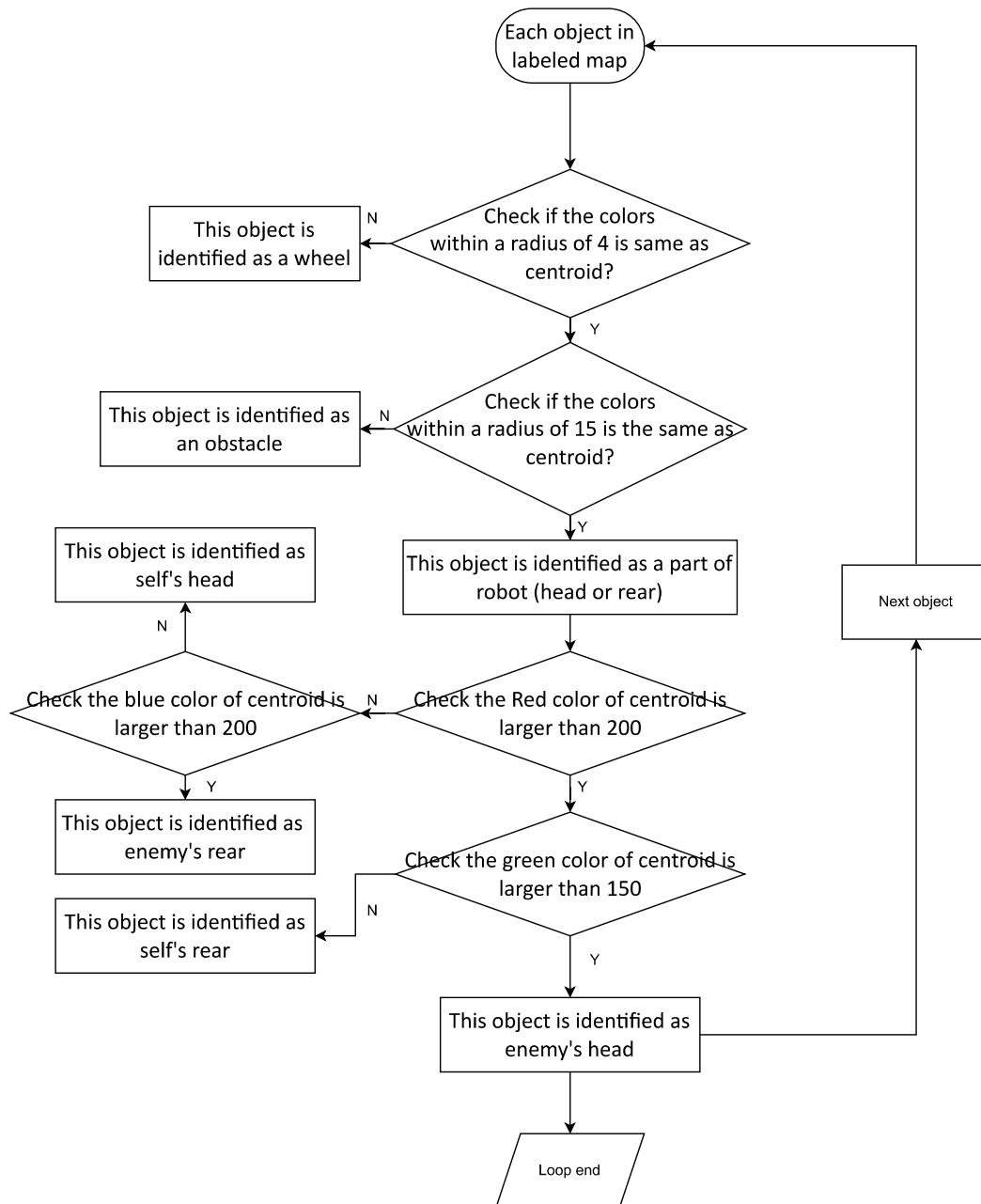


Figure 2.3: Identifying flowchart.

The final result of image processing is shown in figure 2.4

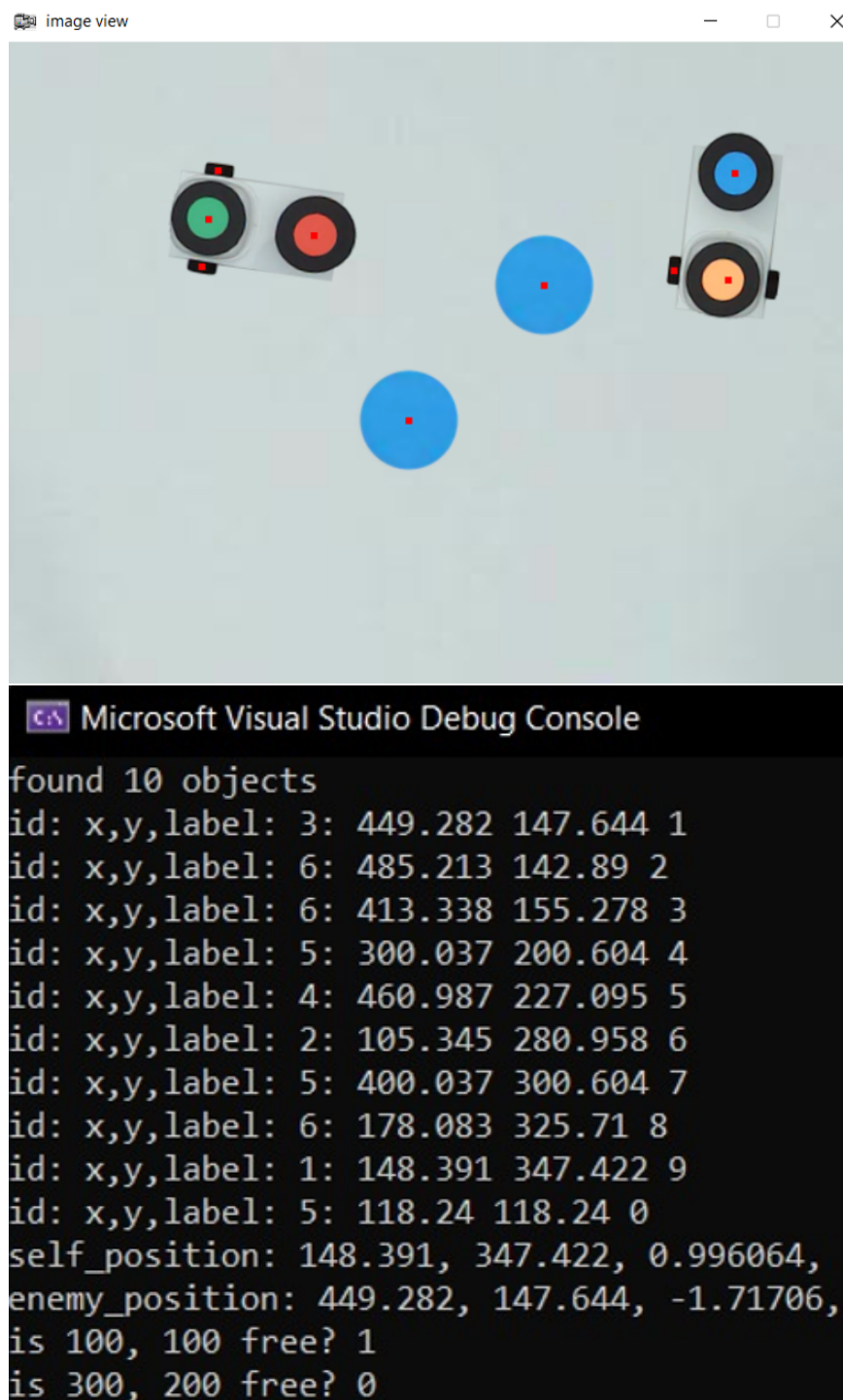


Figure 2.4: Locating all objects.

Chapter 3

Path planning

After getting the position information of car, obstacles, and boundary by image processing, if we want to hide and attack using laser, we really need to control car move from current position to expected position in a special space seeing Figure 3.1, But how to move robot effectively is hard question. Robot path planning is used to find a collision-free sequence of motions (path point) from current position to expected position within a specified environment.

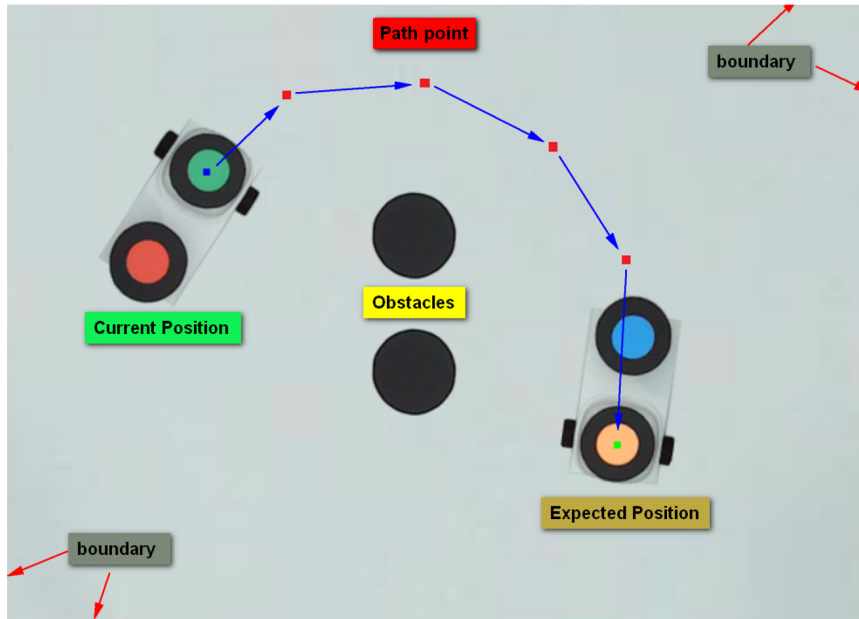


Figure 3.1: How to find the path point.

3.1 The principle and process of ASIA

According to the tasks and requirements, a novel sampling and iterative based path planning algorithm (called Arc Sampling and Iterative Algorithm(ASIA)) is designed to used in this project. ASIA uses four steps to find and renew a new path point.

1. Step 1: Get sampling points

As shown in Figure 3.2, First, we need to build a body coordinate system. The body coordinate

origin is current position, and axis X points the the expected position, and axis Y and axis X intersect at right angles. Second, initialize the value of sampling distance, sampling radian interval and total sampling radian. And get sampling sequence around current position.

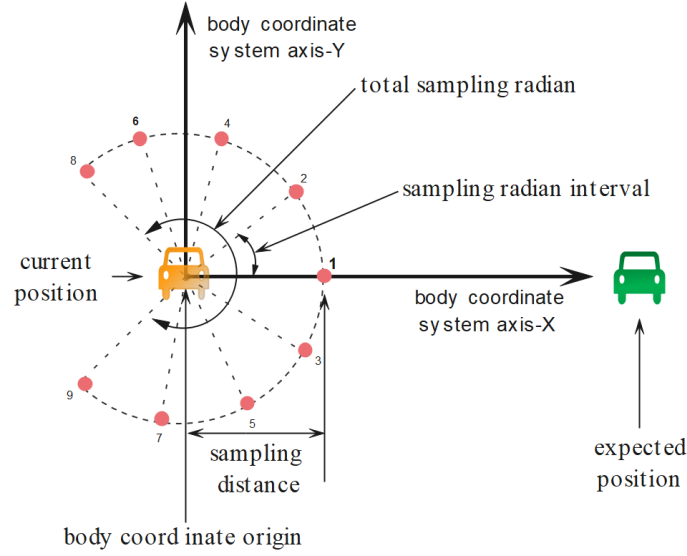


Figure 3.2: How to get sampling points.

2. Step 2: Coordinate transformation

As seen in Figure 3.3, In order to get path point sequence in image coordinate system We need to translate sampling points from body coordinate system to image coordinate system using rotation matrix after getting sampling points.

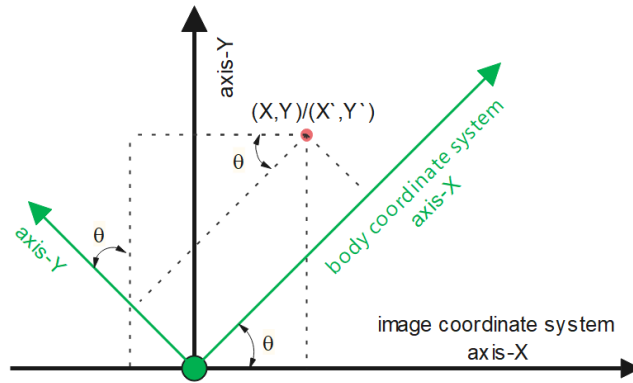


Figure 3.3: coordinate transformation.

According to geometric knowledge in figure 3.3, if the origin position of body coordinate system is (X_0, Y_0) in the mage coordinate system, a translation transform need to be added.

$$\begin{bmatrix} X \\ Y \end{bmatrix} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} X' \\ Y' \end{bmatrix} + \begin{bmatrix} X_0 \\ Y_0 \end{bmatrix} \quad (3.1)$$

3. Step 3: Get path point

As seen in Figure 3.4, Judging whether the sampling points(here we only show sampling points 1-9) is inside the free space sequentially (not coincide with the obstacle and is located within the image boundary). Once the conditions are met , the sampling points will be taken as the path point and renew current position.

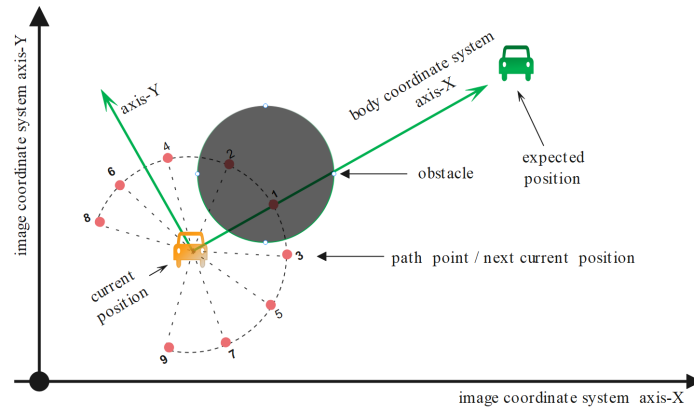


Figure 3.4: how to get path point and renew current position.

4. Step 4: Iterate to get all of the path point

Iterate steps 1 3, until the distance between the current position and the expected position is less the the sampling distance.

3.2 The implement of ASIA

There are two parts codes to implement the ASIA. As seen in Figure 3.5, there are four Sub functions Sampling _ Array, Rotation _ Array(including TF _ X and TF _ Y), Get _ Angle _ Rotation and PathTrack to support ASIA.

```

//part 1-----The Funciton Defination for Path planning-----Starts
// Author : Xiaobo Wu
// Data : 2022.04.11/edition-1
// Function: path planning part include four subfunctions:
//
// 1.SamplingArray: Create Sample point in Body coordinate system;
// 2.RotationArray: Transform Sample point from Body coordinate system to Goble coordinate system
// 3.Get_Angle_Rotation: Get part angle for transform of rotation Array
// 4.PathTrack : When you know the start position(the current position of car) and end/Goal position(the position where you want car to go),You can use this function

//-----1.SamplingArray: Create Sample point from Body coordinate system-----Start
void SamplingArray(float Sampling_points[(270 / 5 + 1) * 3], float Mini_Radian, float S_R){...}
//-----2.RotationArray: Transform Sample point-----
int TF_X(float Reason[2], int Translation_x, float Radian){...}
int TF_Y(float Reason[2], int Translation_y, float Radian){...}
//-----3.Get the angle from Start_A(Current)and Goal B in the map coordinate system-----
float Get_Angle_Rotation(int Start[2], int Goal[2]){...}
//1.part-----The Funciton Defination for Path planning-----End
void PathTrack(int OneViewPoint[2], int A_Global_Start[2], int B_Global_End[2], float Mini_Radian, float S_R, std::vector<object>& objects){...}

```

Figure 3.5: The importance of self-robot direction

Chapter 4

Robot Control

The robot is required to move to any point inside the image frame by controlling the wheels speed. This goal can be achieved by designing a controller to correct for position errors sensed by the camera. To design the controller, the knowledge of the robot's motion is extremely important. The kinematics model of the robot was discussed in section ?? as it is used in this section.

In literature, the commands for the system (controller output) are usually the linear velocity v_c and the angular velocity ω_c . First, a relationship between the controller output and the wheels' motor angular velocity is needed. Solving for the motors' angular velocities gives the following:

$$\omega_r = \frac{1}{R_w} \left(v_c + \frac{D}{2} \omega_c \right) \quad (4.1)$$

$$\omega_l = \frac{1}{R_w} \left(v_c - \frac{D}{2} \omega_c \right) \quad (4.2)$$

Where R_w is the wheel radius. These are called the inverse kinematic equations.

The trajectory is described by a path planning algorithm. This trajectory is broken down into single points (x and y coordinates) to be followed by the robot. Therefore, the controller must bring the robot to these points that are calculated in real time. The design of this controller is described by the block diagram in Figure 4.1.

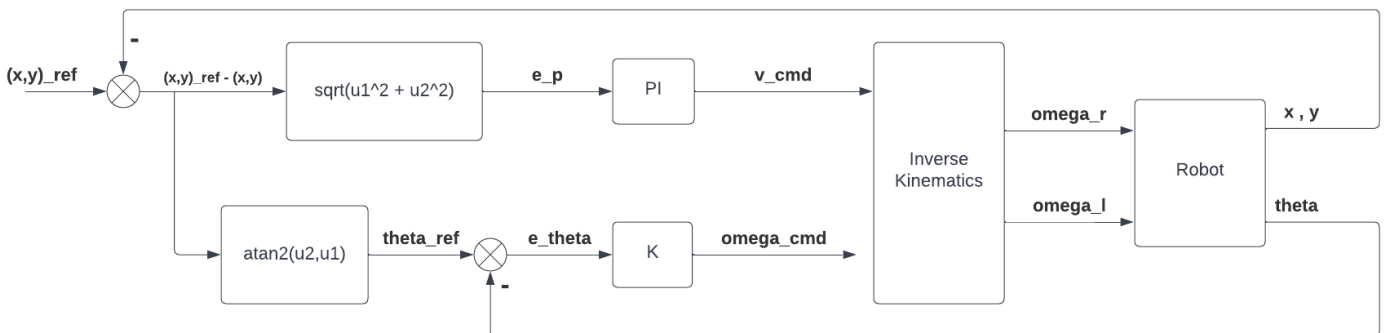


Figure 4.1: Controller block diagram

The reference (x,y) coordinates are given by the path planning algorithm. These coordinates are compared to the position measured by the image processing algorithm to compute an error. The errors are described as:

$$e_p = \sqrt{(x_{ref} - x)^2 + (y_{ref} - y)^2} \quad (4.3)$$

$$e_\theta = \theta_{ref} - \theta \quad (4.4)$$

$$\theta_{ref} = \text{atan2} \left(\frac{y_{ref} - y}{x_{ref} - x} \right) \quad (4.5)$$

There is an error of the distance between the desired point and our measured point, and there is another error that calculates the difference between the robot's orientation and the angle between the robot's x-axis and the distance vector. The first error helps the robot to approach the desired coordinate as best as possible and the second error orients the robot towards the desired coordinates. The position error is fed to a PI controller and the orientation error is fed to a proportional controller to generate the following commands:

$$v_{cmd} = k_p e_p + k_i \int e_p dt \quad (4.6)$$

$$\omega_{cmd} = k_p e_\theta \quad (4.7)$$

The controller gains were tuned for best performance. The robot was tested with multiple reference locations to measure the step response. Some of these results are shown in Figure 4.2. Large distances were input to the robot model to check if there was still a fast response. Most of the time the robot operates with the maximum speed which ensures that the robot moves as fast as possible. It is expected to have a similar speed response for shorter distances.

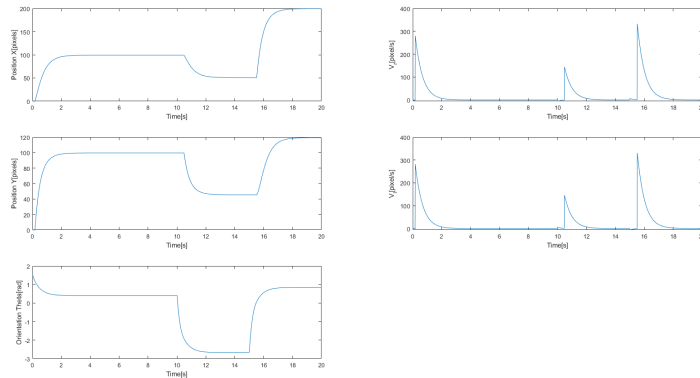


Figure 4.2: Robot response to multiple reference positions

Chapter 5

Hiding and Attacking Strategies

As we all know, there are two tasks everyone robot / program should be able to complete on the vision simulator. First, Your robot chases the opponent robot and tries to hit it with the laser while the other robot tries to avoid getting hit. Second, The opponent robot chases your robot and tries to hit it with the laser while your robot tries to avoid getting hit. We find that obstacles are considered to block the laser so we give the following hiding and attacking strategies.

5.1 Hiding

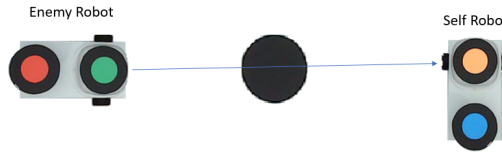


Figure 5.1: an example of positions that satisfies the hiding strategy.

In order to find the nearest point that satisfies the condition, an obstacle between the self and enemy, the foot of perpendicular from current position to the straight-line connecting enemy and obstacles, as shown in figure 5.2

The calculation of the foot of the perpendicular is as follows: We assume that $A = (x_a, y_a)$, $B = (x_b, y_b)$ are position of enemy and obstacle as shown in figure 5.3. $C = (x_c, y_c)$ is the position of self-robot, $O = (x_o, y_o)$ is the foot of perpendicular from point C to line AB .

Since $\overrightarrow{AB} \perp \overrightarrow{CO}$, we have:

$$(x_b - x_a)(x_o - x_c) + (y_b - y_a)(y_o - y_c) = 0 \quad (5.1)$$

Since \overrightarrow{AB} and \overrightarrow{AO} has same direction, we have:

$$\begin{cases} x_o = k(x_b - x_a) + x_a \\ y_o = k(y_b - y_a) + y_a \end{cases} \quad (5.2)$$

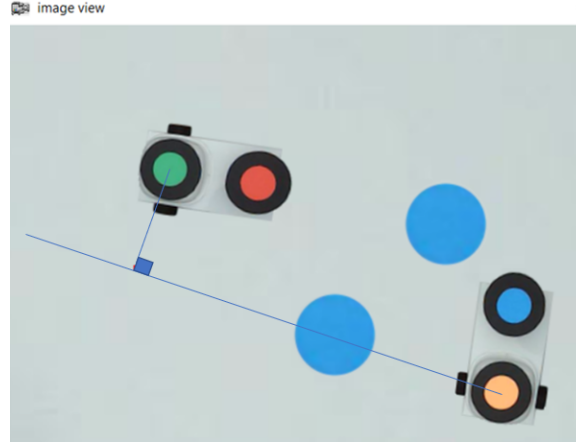


Figure 5.2: the foot of perpendicular from current position to the straight-line connecting enemy and obstacles.

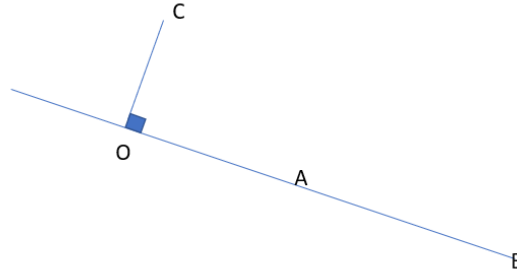


Figure 5.3: The foot of perpendicular.

Substitute x_o and y_o in the second equation to find out k . Then substitute k in the third equation to find out x_o and y_o .

$$k = -\frac{(x_a - x_c)(x_b - x_a) + (y_a - y_c)(y_b - y_a)}{(x_b - x_a)^2 + (y_b - y_a)^2} \quad (5.3)$$

Once we have arrived at the expected position, direction of the self-robot is also important. As shown in figure 5.4 a, if the direction of self-robot is not opposed to the enemy-robot, to run away from the enemy, the self-robot should rotate first, and then move around the obstacle. However, if we keep the direction of self-robot as opposed to the enemy, as shown in figure 5.4 b, we do not need to rotate robot first, we can move robot around the obstacle directly instead.

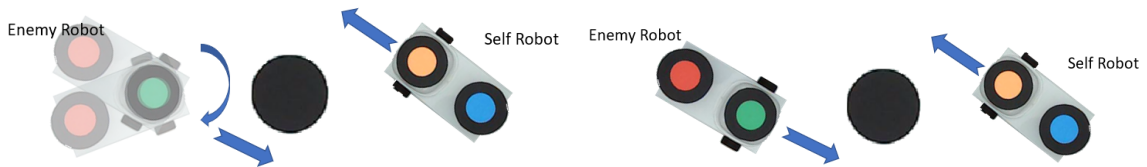


Figure 5.4: The importance of self-robot direction

Therefore, to keep the direction as opposed to the enemy, the expected position also requests a specific theta. For the convenience of calculation, we should stipulate that the theta of robot is between π and $-\pi$.

5.2 Attacking

Since there is only one chance to fire the laser, it should fire until there is no obstacle between two robots. To do so, the attacking robot should chase the other robot. We defined the criteria to fire the laser when:

- The laser is pointing to one of the enemy robot's centroids
- There is a free path (no obstacles) between the laser position and the enemy robot.

Inside the while loop, a series of conditions are programmed to check if the robot can fire the laser. For the first condition, the laser will be tracking the nearest enemy's centroid. Figure 5.5 shows the distances measured by the program. These distances are compared and the smallest one is provided to the laser track algorithm. This function is just a simple adjustment to the laser's servo motor according to the error angle between the laser and the centroid location measured in the image frame. The function will output the error angle for the first condition to be met.

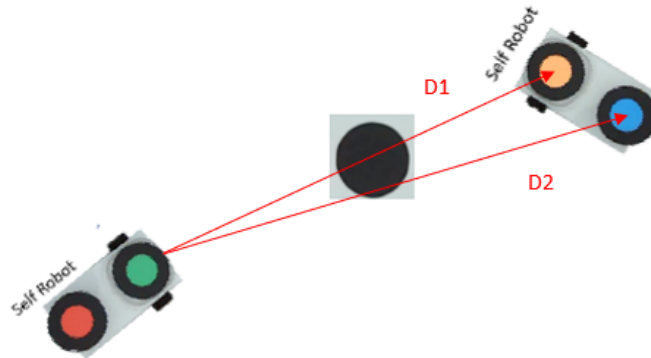


Figure 5.5: Laser tracks the smallest distance to the robot's enemy

The second condition is met when there is no obstacles in the laser's path. Once the nearest centroid is chosen, the `freepath()` function will check if the laser is passing through an obstacle or not. This function keeps track of the pixels that are in between the laser and the enemy's chosen centroid. Then, the function `checkspace()` will tell if the pixel is an obstacle or not.

When condition one and condition two are met, the robot will laser fire to the opponent.

To meet these conditions, the robot has to find a proper location with respect to the enemy. The strategy consists on chasing the enemy's closest centroid. This is done by using the mentioned path planning algorithm which finds the best trajectory while avoiding obstacles. When the described conditions are met, the robot will be able to fire the laser to the enemy.