

# Fundamental of Mobile Robot AUT-710

## Exercise 5

Muhamad Fikri

Widhi Atman

15.3.2024, **SA205**, 10:15 - Finish

# General Plan for Exercises

- Exercise 4: Implementation of model + Basic Control
- Exercise 5: Collision Avoidance with SI model
  - Point obstacle – switching behavior with obstacle avoidance
  - non-point obstacle – switching behavior with wall-following
  - Point obstacles – QP-based controller

10 point

20 point

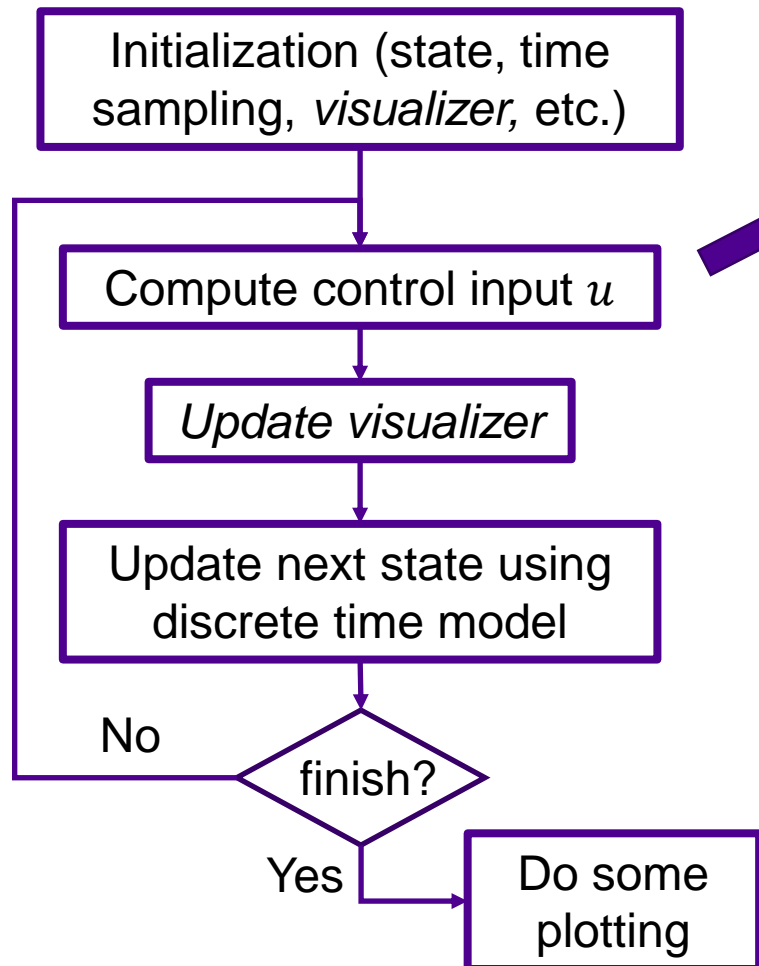
**Deadline: Monday 01.04.2024 at 23:59**

- Exercise 6: Control of Unicycle
- Mini Projects (optional)

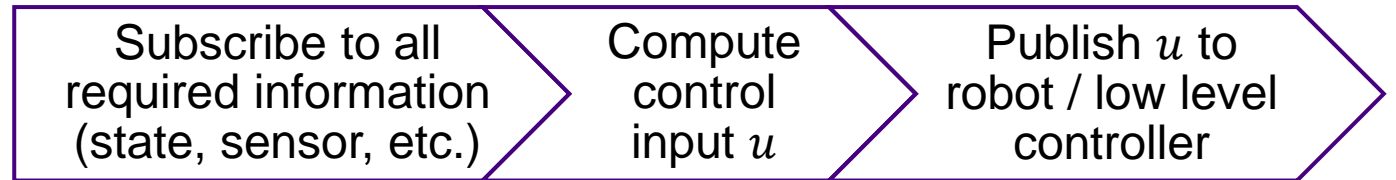
20 point

Bonus up to 20 point

# Flowchart of Simulator



*If you are interested in implementing this to ROS*



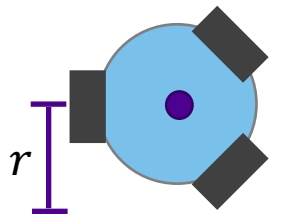
## Specifications (for Exercise 5)

Time sampling  $T = 10ms$

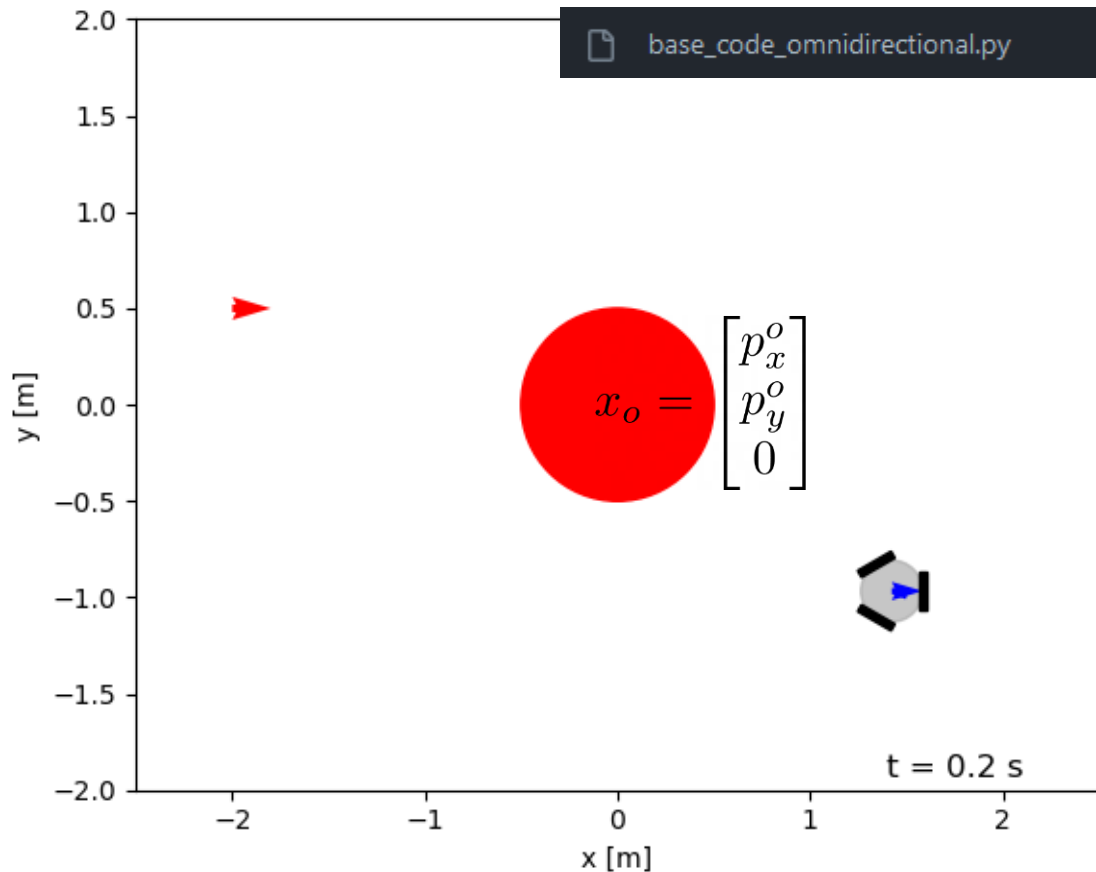
Robot's radius =  $0.21\text{ m}$

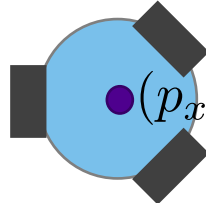
Max translational vel.  $(v_x^2 + v_y^2)^{\frac{1}{2}} = 0.5\text{ m/s}$

Max rotational vel.  $(|\omega|) = 5\text{ rad/s}$



# Exercise 5.1 – Scenario





$$x = \begin{bmatrix} p_x \\ p_y \\ \theta \end{bmatrix} \quad u = \begin{bmatrix} v_x \\ v_y \\ \omega \end{bmatrix}$$

**Model:** omnidirectional mobile robot (**single-integrator model**)

**Initial Position:**  $x[0] = [1.5 \quad -1 \quad *]^T$

**Goal:** static at  $x^d = [-2 \quad 0.5 \quad 0]^T$ .

\* Can be any orientation at goal position

**Key Scenario:**

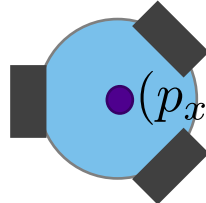
- We assume the robot/controller can identify a **circular obstacle** (centroid and radius) once it is near. Here, the obstacle is centered at  $p_x^o = 0$ ,  $p_y^o = 0$  with radius  $0.5m$ .

**Control Objective:**

- Reach the goal while avoiding contact/collision with obstacle

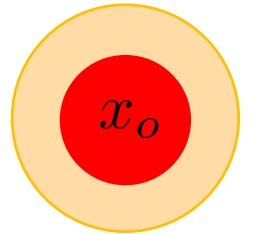
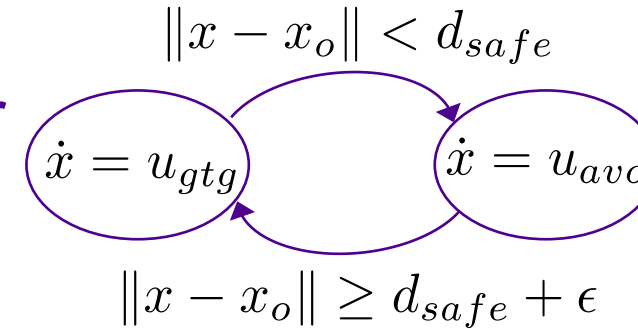
# Exercise 5.1 – Task

5 points



$$x = \begin{bmatrix} p_x \\ p_y \\ \theta \end{bmatrix} \quad u = \begin{bmatrix} v_x \\ v_y \\ \omega \end{bmatrix}$$

By taking account of the **robot's size and limitation**, design and implement the following **switched controller**



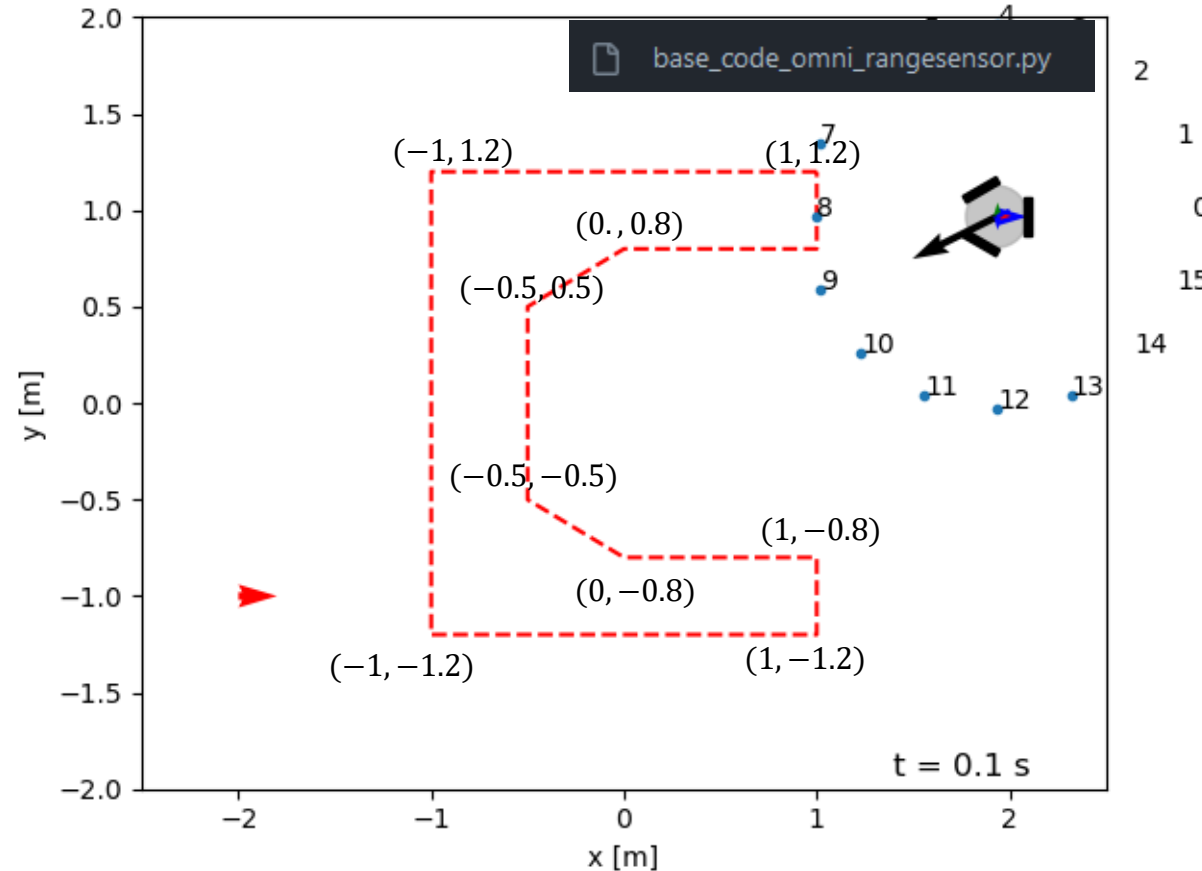
Describe your approach in designing  $u_{gtg}$ ,  $u_{avo}$ ,  $d_{safe}$ , and  $\epsilon$ , as well as your observation on the resulting controller.

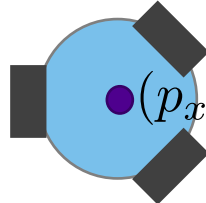
Show the result by plotting:

- *time series* of control input  $u$  and  $(v_x^2 + v_y^2)^{0.5}$
- *time series* of error  $(x^d - x)$ ,
- *time series* of distance to obstacle  $\|x - x_o\|$
- *time series* of state trajectory  $x$  vs  $x^d$ , and
- XY **trajectory** of the robot (or final snapshot of the simulator).

Deadlock issue: Try to set the initial position to  $x[0] = [2 \quad -0.5 \quad 0]^T$  and see what happen.

# Exercise 5.2 – Scenario





$$x = \begin{bmatrix} p_x \\ p_y \\ \theta \end{bmatrix} \quad u = \begin{bmatrix} v_x \\ v_y \\ \omega \end{bmatrix}$$

**Model:** omnidirectional mobile robot  
(**single-integrator model**)

**Initial Position:**  $x[0] = [2 \quad 1 \quad 0]^T$

**Goal:** static at  $x^d = [-2 \quad -1 \quad *]^T$ .

\* Can be any orientation at goal position

**Key Scenario:**

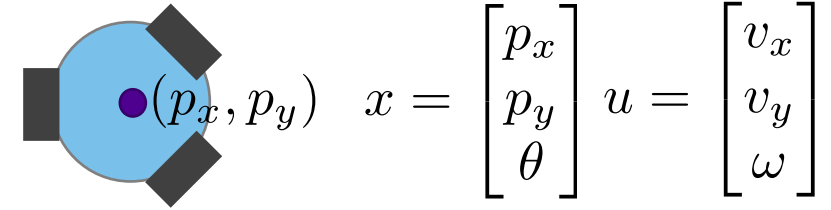
- An obstacle presents in the field, but **unknown** to the robot/controller.
- The obstacle is detected by the reading from range sensor ( $< 1 \text{ m}$ )  
*Assume accurate readings from sensors.*

**Control Objective:**

- Reach the goal while avoiding contact/collision with obstacle

# Exercise 5.2 – Task

10 points



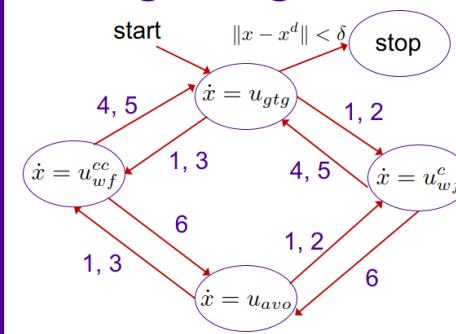
By taking account of the **robot's size and limitation**, design and implement **wall-following behavior** to your switching controller in 5.1.

Describe your approach in designing  $u_{wf}^c$ ,  $u_{wf}^{cc}$ , and computing  $x_o$  (and  $u_{avo}$ ) from sensor readings as well as your observations on the resulting controller.

Show the result by plotting:

- *time series* of control input  $u$  and  $(v_x^2 + v_y^2)^{0.5}$
- *time series* of error  $(x^d - x)$ ,
- *time series* of *minimum reading distance* from the sensor
- *time series* of state trajectory  $x$  vs  $x^d$ , and
- XY **trajectory** of the robot (or final snapshot of the simulator).

## Putting all together



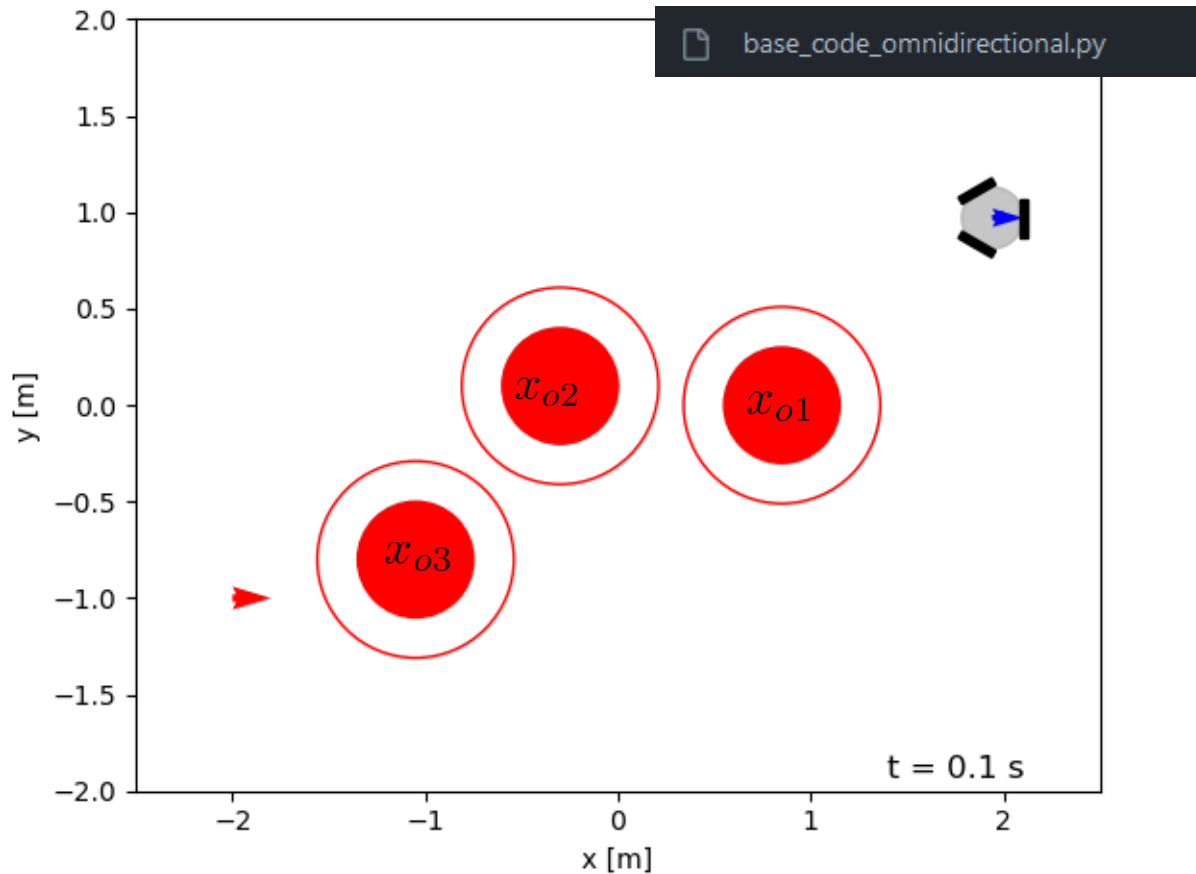
Conditions for switching:

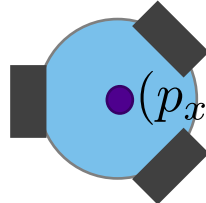
- (1)  $d_{safe} - \epsilon \leq \|x - x_o\| \leq d_{safe} + \epsilon$
- (2)  $u_{gtg}^T u_{wf}^c > 0$
- (3)  $u_{gtg}^T u_{wf}^{cc} > 0$
- (4)  $u_{avo}^T u_{gtg} > 0$
- (5)  $\|x(t) - x^d\| < \|x(t_s) - x^d\|$
- (6)  $\|x - x_o\| < d_{safe} - \epsilon$

## IMPORTANT TIPS:

- Use visualization to help debug (e.g., the  $u_{gtg}$ ,  $u_{avo}$ ,  $u_{wf}^c$ ,  $u_{wf}^{cc}$ , etc.)
- *Print* every time the state changes

# Exercise 5.3 – Scenario





$$x = \begin{bmatrix} p_x \\ p_y \\ \theta \end{bmatrix} \quad u = \begin{bmatrix} v_x \\ v_y \\ \omega \end{bmatrix}$$

**Model:** omnidirectional mobile robot (**single-integrator model**)

**Initial Position:**  $x[0] = [2 \quad 1 \quad 0]^T$

**Goal:** static at  $x^d = [-2 \quad -1 \quad *]^T$ .

\* Can be any orientation at goal position

## Key Scenario:

- We assume the robot/controller can identify a **circular obstacle** (centroid and radius) once it is near. Here, 3 obstacle presents:
  1. centered at  $p_x^{o1} = 0.85$ ,  $p_y^{o1} = 0$  with radius  $0.3m$ .
  2. centered at  $p_x^{o2} = -0.3$ ,  $p_y^{o2} = 0.1$  with radius  $0.3m$ .
  3. centered at  $p_x^{o3} = -1.05$ ,  $p_y^{o3} = -0.8$  with radius  $0.3m$ .

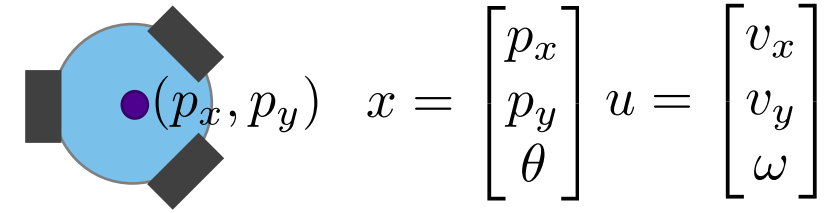
## Control Objective:

- Reach the goal while avoiding contact/collision with obstacle



# Exercise 5.3

5 point



By taking account of the robot's limitation, implement the **QP-based controller** as follows

$$u = \arg \min_{u^*} \|u_{gtg} - u^*\|^2$$

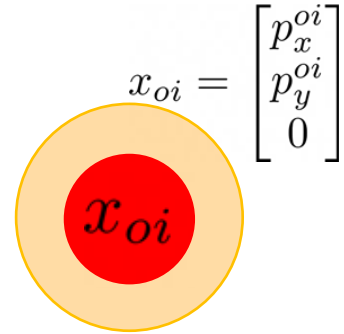
$$\text{s.t.} \quad \left(\frac{\partial h_{o1}}{\partial x}\right)^T u^* \geq -\gamma(h_{o1}(x))$$

$$\left(\frac{\partial h_{o2}}{\partial x}\right)^T u^* \geq -\gamma(h_{o2}(x))$$

$$\left(\frac{\partial h_{o3}}{\partial x}\right)^T u^* \geq -\gamma(h_{o3}(x))$$

$$\text{with } h_{oi} = \left\| \begin{bmatrix} p_x \\ p_y \end{bmatrix} - \begin{bmatrix} p_x^{oi} \\ p_y^{oi} \end{bmatrix} \right\|^2 - R_{si}^2$$

Use  $R_{si} = 0.51$  for all obstacles.



**NOTE:**  $u_{gtg}$  still need to comply with the robot's max speed. You can use  $u_{gtg}$  from 5.1.

Then, test the controller for  $\gamma(h) = 0.2h$ ,  $\gamma(h) = 10h$ , and  $\gamma(h) = 10h^3$  and describe your observation on what does the variation affects.

Show the result by comparing the plot for each  $\gamma$  function variation via:

- Time series comparison of control input  $u_{gtg}$  vs  $u$  (plot only  $v_x$  and  $v_y$ )
- Time series comparison of function  $h_{o1}$ ,  $h_{o2}$ , and  $h_{o3}$
- Comparison of XY **trajectory** of the robot

# How to show obstacle on Simulator?

Plot using the axis object in the `sim_mobile_robot` class → `sim_visualizer.ax`

**Option 1:** using patch in matplotlib (for simple shape: circle, rectangle, etc.)

```
if IS_SHOWING_2DVISUALIZATION: # Initialize Plot
    sim_visualizer = sim_mobile_robot( 'omnidirectional' ) # Omnidirectional Icon
    #sim_visualizer = sim_mobile_robot( 'unicycle' ) # Unicycle Icon
    sim_visualizer.set_field( field_x, field_y ) # set plot area
    sim_visualizer.show_goal(desired_state)
    sim_visualizer.ax.add_patch( plt.Circle( (0, 0), 0.5, color='r' ) )
    sim_visualizer.ax.add_patch( plt.Circle( (0, 0), d_safe, color='r', fill=False) )
    sim_visualizer.ax.add_patch( plt.Circle( (0, 0), d_safe + eps, color='g', fill=False) )
```

*Example for 5.1*

Draw (full) red circle for obstacle  
 Draw empty red circle for  $d_{safe}$   
 Draw empty green circle for  $d_{safe} + \epsilon$

**Option 2:** by defining obstacle's vertices and use line plot

```
if IS_SHOWING_2DVISUALIZATION: # Initialize Plot
    sim_visualizer = sim_mobile_robot( 'omnidirectional' ) # Omnidirectional Icon
    #sim_visualizer = sim_mobile_robot( 'unicycle' ) # Unicycle Icon
    sim_visualizer.set_field( field_x, field_y ) # set plot area
    sim_visualizer.show_goal(desired_state)

    # Display the obstacle
    obst_vertices = np.array( [ [-1., -1.2], [1., -1.2], [1., 1.2], [-1., 1.2], \
                                [-1., 0.8], [0., 0.8], [0.5, 0.5], [0.5, -0.5], [0., -0.8], [-1., -0.8], [-1., -1.2] ] )
    sim_visualizer.ax.plot( obst_vertices[:,0], obst_vertices[:,1], '--r' )
```

*Example for 5.2*

# How to implement QP for 5.3?

$$u = \arg \min_{u^*} \|u_{gtg} - u^*\|^2$$

$$\text{s.t.} \quad \left(\frac{\partial h_{o1}}{\partial x}\right)^T u^* \geq -\gamma(h_{o1}(x))$$

$$\left(\frac{\partial h_{o2}}{\partial x}\right)^T u^* \geq -\gamma(h_{o2}(x))$$

$$\left(\frac{\partial h_{o3}}{\partial x}\right)^T u^* \geq -\gamma(h_{o3}(x))$$

Refer to lecture 10  
on 22.3.2023



$$\min_z \frac{1}{2} z^T Q z + c^T z$$

$$\text{s.t.} \quad H z \preceq b$$

`import cvxopt` Require **cvxopt** python package

```
# QP-based controller
# -----
Q_mat = 2 * cvxopt.matrix( np.eye(2), tc='d')
c_mat = -2 * cvxopt.matrix( u_gtg[:2], tc='d')

# Fill H and b based on the specification afterwards
# --> row is number of constraints / specification
H = np.zeros([row, 2]) # TODO
b = np.zeros([row, 1]) # TODO

# Resize the H and b into appropriate matrix for optimization
H_mat = cvxopt.matrix( H, tc='d')
b_mat = cvxopt.matrix( b, tc='d')

# Solving Optimization
cvxopt.solvers.options['show_progress'] = False
sol = cvxopt.solvers.qp(Q_mat, c_mat, H_mat, b_mat, verbose=False)

current_input = np.array([sol['x'][0], sol['x'][1], 0])
```

→  $z = \begin{bmatrix} v_x \\ v_y \end{bmatrix}$

→ Change these  
two lines to the  
appropriate  
values

*NOTE: These code only compute  
the linear velocity (x and y) which  
is sufficient for 5.3.*

# Question?

- Consult them via
  - Exercise sessions on 15.3.2024, 22.3.2024, and 29.3.2024
  - Teams channel