# Task-Priority kinematic control Obstacle avoidance and Joint limit

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March 16, 2025

# Introduction

This lab report presents the implementation of a priority control task for a 3-degree-of-freedom (3-DOF) planar robot Fig 1. The focus is on integrating an obstacle avoidance task and a joint limiting task while the robot attempts to reach a specified 2D target point. A hierarchical control approach is used to ensure that critical constraints, such as collision avoidance and joint limit enforcement, take priority while still allowing the robot to achieve its objective. The lower-priority task operates within the null space left by the higher-priority tasks, ensuring that it does not interfere with more critical constraints.

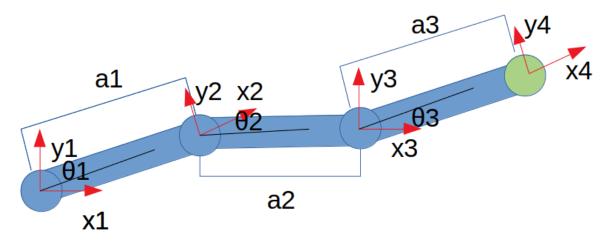


Fig 1. DH in the robot.

#### Obstacle avoidance:

Activating the obstacle avoidance task at all times can lead to inefficient robot movements or, in some cases, prevent the robot from reaching its target due to task prioritization constraints. To prevent this issue, the obstacle avoidance task is only activated within a specific range, ensuring that it affects the robot's motion only when necessary. This activation range is controlled using a hysteresis-based logic, which includes both an activation threshold and a deactivation threshold. The task is triggered when the robot comes too close to an obstacle and deactivates once it moves a safe distance away. This prevents frequent, unnecessary activations that could lead to erratic movements (chatter).

The activation of the obstacle avoidance task is determined using a variable,  $\sigma$ , which represents the direction in which the robot should move to avoid the obstacle. This value is calculated as the vector from the end effector to the center of the goal, divided by its norm, resulting in a unit vector pointing away from the obstacle. The activation and deactivation thresholds are set based on the distance from the center of the obstacle, tuning these parameters lead to smoother trajectories. The Jacobian used for this task is the same as the one for the 2D position control, as both tasks control the x and y coordinates of the robot's last joint.

Due to the way the obstacle avoidance task is implemented, it is only performed by the last joint (end effector, EE), and the rest of the arm is not taken into consideration. This limitation can be addressed by adding multiple obstacles along the arm's trajectory. However, as presented in this report, the current approach is specifically suited for SCARA-type robots, where the focus is on controlling the end effector's position. As demonstrated in Figure 2, the robot successfully reaches the target position, avoiding all obstacles, unless the randomly generated target is located inside an obstacle.

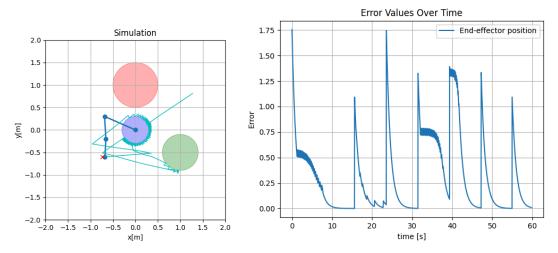


Fig 2. Reaching a 2d point avoiding obstacles.

# Joint limit:

Similarly to the obstacle avoidance task, an activation function is implemented for both the upper and lower joint limits. This function includes activation and deactivation thresholds to ensure the task only engages when necessary. The activation logic is based on comparing the current joint position ( $\sigma$ ) with predefined upper and lower limits, incorporating both activation and deactivation thresholds. If the joint position exceeds the lower limit minus the activation threshold, the task is activated in the negative direction, if the joint position is below the upper limit plus the activation threshold, the task is activated in the positive direction.

The error value is always either 1, 0, or -1, depending on the sign of the activation function, which is determined directly by whether the joint position exceeds the specified upper or lower limits. If the joint moves away from the limit beyond the deactivation threshold, the task is deactivated, ensuring the robot operates within safe bounds and prevents unnecessary or excessive corrections.

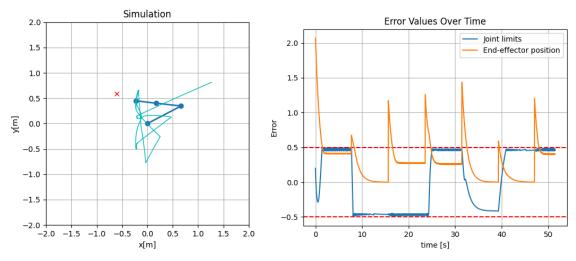


Fig 3. Reaching a 2d point limiting joint 1 angle.

As can be seen in Fig 3, the joint limit control works well, preventing the first joint from exceeding the set limits. This limitation significantly reduces the robot's operative space. As a result, the robot is unable to reach the desired position in several instances. The red lines in the figure represent the upper and lower joint limits, and it is evident that the joint (represented by the blue line) does not exceed these limits. This constraint ensures safety but also demonstrates the trade-off between maintaining joint limits and limiting the robot's range of motion, which can impact its ability to reach certain target positions. It can be clearly seen that when the joint reaches the limit the end effector position error starts and asymptotic behavior.

### Code:

## **Control Loop**

```
from lab4 robotics import * # Includes numpy import
import matplotlib.pyplot as plt
import matplotlib.animation as anim
import matplotlib.patches as patch
import time
d = np.zeros(3)
theta = np.array([0.2,0.5,0.4]).reshape(1,3)[0]
alpha = np.zeros(3)
a = np.array([0.75, 0.5, 0.4]).reshape(1,3)[0]
revolute = [True,True,True]
robot = Manipulator(d, theta, a, alpha, revolute)
obstaclePos 1 = np.array([0.0, 1.0]).reshape(2,1)
obstacleR 1 = 0.5
obstaclePos_2 = np.array([0.0, 0.0]).reshape(2,1)
obstacleR 2 = 0.3
obstaclePos 3 = np.array([1.0, -0.5]).reshape(2,1)
obstacleR 3 = 0.4
obstacle vec = np.array([obstaclePos 1, obstaclePos 2, obstaclePos 3])
obstacle_r = np.array([obstacleR_1, obstacleR_2, obstacleR_3])
obstacle color = ['red', 'blue', 'green']
limits = np.array([0.5, -0.5])
tasks = [
       Obstacle2D("Obstacle avoidance", obstaclePos 1, np.array([obstacleR 1,
obstacleR 1+0.05]), robot),
       Obstacle2D("Obstacle avoidance", obstaclePos 2, np.array([obstacleR 2,
obstacleR_2+0.05]), robot),
```

```
Obstacle2D("Obstacle avoidance", obstaclePos_3, np.array([obstacleR_3,
obstacleR 3+0.05]), robot),
dt = 1.0/60.0
Tt = 1000
tt = np.arange(0, Tt, dt)
start time = time.time()  # Start time
fig = plt.figure()
ax = fig.add subplot(111, autoscale on=False, xlim=(-2, 2), ylim=(-2,2))
ax.set title('Simulation')
ax.set aspect('equal')
ax.grid()
ax.set xlabel('x[m]')
ax.set ylabel('y[m]')
for i in range(len(obstacle vec)):
  ax.add patch(patch.Circle(obstacle pos.flatten(), obstacle rad,
color=obstacle color[i], alpha=0.3))
line, = ax.plot([], [], 'o-', lw=2) # Robot structure
path, = ax.plot([], [], 'c-', lw=1) # End-effector path
PPx = []
PPy = []
def init():
  return line, path, point
def simulate(t):
```

```
null space = np.eye(robot.dof)
  dq = np.zeros(robot.dof).reshape(-1, 1)
      i.update(robot)
      if i.isActive():
          dq += Jbar inv @ ((i.getK()@i.getError()-J@dq) + i.ff) # calculate
quasi-velocities with null-space tasks execution
  current time = time.time()
  robot.update(dq, dt)
  PP = robot.drawing()
  PPx.append(PP[0,-1])
  PPy.append(PP[1,-1])
  path.set data(PPx, PPy)
  point.set data(tasks[-1].getDesired()[0], tasks[-1].getDesired()[1])
animation = anim.FuncAnimation(fig, simulate, np.arange(0, Tt, dt),
                              interval=10, blit=True, init_func=init, repeat=True)
```

```
plt.show()

# Plot errors
plt.figure()
for i in tasks:
    if type(i) is Configuration2D:
        plt.plot(tt[:len(i.erroVec[0])], i.erroVec[0], label='Position Error')
        plt.plot(tt[:len(i.erroVec[1])], i.erroVec[1], label='Angular Error')
elif type(i) is Obstacle2D:
        continue
elif type(i) is JointLimit2D:
        plt.plot(tt[:len(robot.story)], robot.story, label=i.name)
else:
        plt.plot(tt[:len(i.erroVec)], i.erroVec, label=i.name)

# paint line indicating limit of the angles
plt.axhline(y=limits[0], color='r', linestyle='--')
plt.axhline(y=limits[1], color='r', linestyle='--')
plt.title('Error Values Over Time')
plt.title('Error Values Over Time')
plt.ylabel('Error')
plt.grid()
plt.legend()
plt.legend()
```

#### Task Classes

```
Oi = T[i][0:3,3] # Extract o
    J[:,i] = np.concatenate((np.cross(z, O - Oi), z)) # Build the jacobian
self.d = d
self.theta = theta
self.a = a
self.q = np.array(theta).reshape(-1, 1)
```

```
self.update(0.0, 0.0)
def update(self, dq, dt):
   self.story.append(self.q[0][0])
   for i in range(len(self.revolute)):
def drawing(self):
   return robotPoints2D(self.T)
```

```
return self.q[joint]
def getDOF(self):
    return self.T[link]
    return np.array(np.arctan2(linkT[1,0], linkT[0,0])).reshape((1,1))
   self.sigma_d = desired # desired sigma
   self.k = None
def getFF(self):
def setFF(self, ff):
```

```
return self.k
def update(self, robot):
   return self.err
```

```
class Position2D(Task):
      self.J = np.zeros((2,robot.getDOF()))
      self.setK(np.eye(2))
  def update(self, robot: Manipulator):
       self.J = robot.getLINKJacobian(self.link)[:2,:].reshape((2,self.link))
      self.err = np.array(self.getDesired() -
robot.getLinkTransform(self.link)[0:2,3].reshape((2,1)))  # Update task error
       self.erroVec.append(np.linalg.norm(self.err))
  def setRandomDesired(self):
      random = (np.random.rand(2,1)*2-1).reshape((2,1))
      self.setDesired(random)
      self.J = np.zeros((1,robot.getDOF()))# Initialize with proper dimensions
      self.setK(np.eye(1))
      self.setFF(np.zeros((1,1)))
  def update(self, robot: Manipulator):
       self.J = robot.getLINKJacobian(self.link)[5,:].reshape((1,self.link)) #
      self.J = np.pad(self.J, (0, robot.dof - self.link), mode='constant',
```

```
current_sigma = np.array(np.arctan2(current_transform[1,0],
current transform[0,0])).reshape((1,1)) # Compute current sigma
      print('current_sigma:',current_sigma)
      self.err = wrapangle(self.getDesired() - current sigma.reshape((1,1))) # Update
      self.erroVec.append(self.err[0])
      self.setDesired( (np.random.rand(1,1)*2*np.pi-np.pi).reshape((1,1)))
      self.setDesired(np.array([np.pi]).reshape(1,1))
      self.setK(np.eye(3))
      self.setFF(np.zeros((3,1)))
  def update(self, robot: Manipulator):
      positionJacobian =
robot.getLINKJacobian(self.link)[:2,:].reshape((2,self.link))
      positionJacobian = np.hstack((positionJacobian, np.zeros((2, robot.dof -
self.link))))
      orientationJacobian =
robot.getLINKJacobian(self.link)[5,:].reshape((1,self.link))  # Update task Jacobian
self.link))))
current transform[0,0]) # Compute current sigma angle
      current sigma pos = current transform[0:2,3] # Compute current sigma position
```

```
error_pos = self.getDesired()[0:2] - current_sigma_pos.reshape((2,1)) # Compute
      error_angle = self.getDesired()[2] - current_sigma_angle
      print ("angular error: ", error angle)
      self.err = np.array([error_pos[0], error_pos[1], error_angle]).reshape((3,1)) #
      self.erroVec[0].append(np.linalg.norm(error pos))
      self.erroVec[1].append(error angle[0])
  def setRandomDesired(self):
       self.setDesired(np.array([np.random.rand(1,1)*2-1,np.random.rand(1,1)*2-1,
np.random.rand(1,1)*2*np.pi-np.pi]).reshape((3,1)))
      self.setK(np.eye(1))
      self.setFF(np.zeros((1,1)))
  def update(self, robot: Manipulator):
      self.J = np.pad(self.J, (0, robot.dof - self.link), mode='constant',
constant values=0)
      self.err = wrapangle(self.getDesired() - current sigma.reshape((1,1))) # Update
      print ("angular error: ", self.err)
      self.erroVec.append(self.err[0])
  def setRandomDesired(self):
      self.setDesired( (np.random.rand(1,1)*2*np.pi-np.pi).reshape((1,1)))
```

```
class Obstacle2D(Task):
      self.active = False
      self.setK(np.eye(2))
      self.setFF(np.zeros((2,1)))
  def update(self, robot: Manipulator):
       self.J = robot.getEEJacobian()[:2,:].reshape((2,robot.dof))  # Update task
      current sigma = robot.getEETransform()[:2,3].reshape((2,1)) - self.position#g
      self.activate(current sigma)
      print("current_sigma: ",current_sigma)
      self.err = current_sigma/np.linalg.norm(current_sigma)
      self.limits = limits
      self.setK(np.eye(1))
      self.setFF(np.zeros((1,1)))
  def update(self, robot: Manipulator):
      self.J = robot.getLINKJacobian(self.link)[5,:].reshape((1,self.link))
```

```
self.J = np.pad(self.J, (0, robot.dof - self.link), mode='constant',
constant_values=0)
    current_sigma = robot.getLinkOrientation(self.link) # Compute current sigma
    self.activate(current_sigma)
    print('current_sigma:',current_sigma)
    self.err = np.array([1*self.active]) # Update task error
    print ("angular error: ", self.err)
    self.erroVec.append(self.err)
    pass # to remove

def activate (self, angle):
    if self.active == 0 and angle >= self.limits[0] - self.activation_tresh:
        self.active = 1
    elif self.active == 0 and angle <= self.limits[1] + self.activation_tresh:
        self.active == -1 and angle <= self.limits[0] - self.deactivation_tresh:
        self.active == 0
    elif self.active == 1 and angle >= self.limits[1] + self.deactivation_tresh:
        self.active == 0

def isActive(self):
    return abs(self.active)

def setRandomDesired(self):
    self.setDesired( (np.random.rand(1,1)*2*np.pi-np.pi).reshape((1,1)))
    pass
```