

Students:

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# Lab4: Task-Priority kinematic control

## Introduction

This lab session is focused on Task-Priority algorithm in its recursive form, to allow for an arbitrary hierarchy of tasks. The implementation is split into two parts. The first part is the definition of different tasks as Python classes and the second part is the recursive TP itself, with a simulation on a 3-link planar manipulator.

## Methodology

### 1 Exercise 01

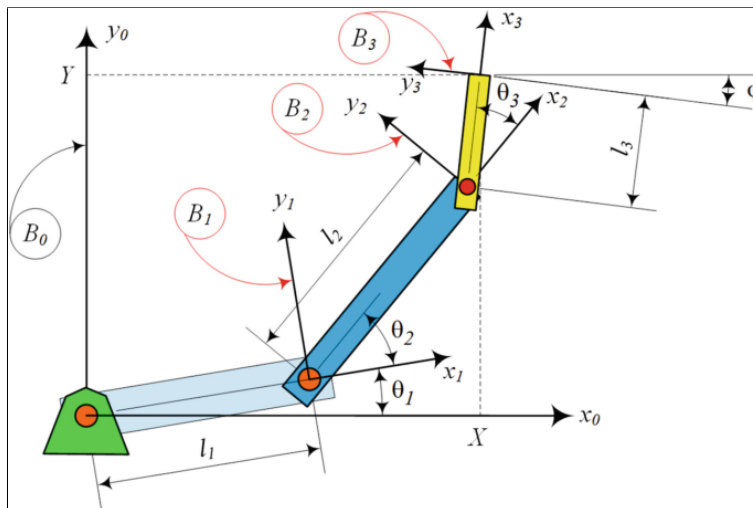


Figure 1: Three links manipulator model, including DH parameters and coordinate systems

Link	$d$	$\theta$	$a$	$\alpha$	Home
1	0	$\theta_1$	$a_1$	0	0
2	0	$\theta_2$	$a_2$	0	0
3	0	$\theta_3$	$a_3$	0	0

Table 1: Denavit-Hartenberg table of the three links manipulator

Denavit-Hartenberg formulation which is used to compute transformation matrix between coordinate systems:

$$T_n^{n-1} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & d_i \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos(\theta_i) & -\sin(\theta_i) & 0 & 0 \\ \sin(\theta_i) & \cos(\theta_i) & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & a_i \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos(\alpha_i) & -\sin(\alpha_i) & 0 \\ 0 & \sin(\alpha_i) & \cos(\alpha_i) & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad (1)$$

$$T_n^0(q) = T_1^0 T_2^1 T_3^2 \dots T_n^{n-1} = \begin{bmatrix} R_n^0(q) & O_n^0(q) \\ 0 & 1 \end{bmatrix} \quad (2)$$

Implementation on python code:

## 1.1 Definition of parameters of the simulation

Listing 1: Robot definition (3 revolute joint planar manipulator)

```
1 # Robot definition (3 revolute joint planar manipulator)
2 d = np.zeros(3) # displacement along Z-axis
3 q = np.array([0.2, 0.5, 0.2]) # rotation around Z-axis (theta)
4 a = np.array([0.75, 0.5, 0.3]) # displacement along X-axis
5 alpha = np.zeros(3) # rotation around X-axis
6 revolute = [True, True, True] # flags specifying the type of joints
7 n_DoF = len(revolute) # Number of Degree of Freedom
8 dq_max = np.array([3, 3, 3]) # The maximum joint velocity limit
9 # Setting desired position of end-effector to the current one
10 sigma_d = np.array([1.0, 1.0])
```

## 1.2 Implementation of subclasses of the base Task class

Listing 2: Position of the end-effector (2D)

```
1 '''
2     Subclass of Task, representing the 2D position task.
3 '''
4 class Position2D(Task):
5     def __init__(self, name, desired, robot: Manipulator):
6         super().__init__(name, desired, robot)
7         self.J = np.zeros((2, robot.getDOF()))
8         self.err = np.zeros((2, 1))
9
10     def update(self, robot: Manipulator, dt: float, newDesired):
```

```

11     DoF = robot.getDOF()
12     # Update Jacobean matrix - task Jacobian
13     self.J = robot.getLinkJacobian(self.link_index)[[0,1]].reshape(
self.task_dim,DoF)
14     # Update task error
15     self.err = self.getDesired() - robot.getJointPos2D(self.
link_index)
16     # Compute feed-forward velocity
17     self.ffVel = (newDesired - self.getDesired()) / dt
18     # Set new desired
19     self.setDesired(newDesired)
20     return True

```

### Listing 3: Orientation of the end-effector (2D)

```

1 '''
2     Subclass of Task, representing the 2D orientation task.
3 '''
4 class Orientation2D(Task):
5     def __init__(self, name, desired, robot: Manipulator):
6         super().__init__(name, desired, robot)
7         self.J = np.zeros((1, robot.getDOF()))
8         self.err = np.zeros((1, 1))
9
10    def update(self, robot: Manipulator, dt: float, newDesired):
11        DoF = robot.getDOF()
12        # Update Jacobean matrix - task Jacobian
13        self.J = robot.getLinkJacobian(self.link_index)[[5]].reshape(
self.task_dim,DoF)
14        # Update task error
15        self.err = self.getDesired() - robot.getJointOrientation2D(self.
link_index)
16        # Compute feed-forward velocity
17        self.ffVel = (newDesired - self.getDesired()) / dt
18        # Set new desired
19        self.setDesired(newDesired)
20        return True

```

### Listing 4: Configuration of end-effector (2D)

```

1 '''
2     Subclass of Task, representing the 2D configuration task.
3 '''
4 class Configuration2D(Task):
5     def __init__(self, name, desired, robot: Manipulator):

```

```

6         super().__init__(name, desired, robot)
7         self.J = np.zeros((6, robot.getDOF())) #
           Initialize with proper dimensions
8         self.err = np.zeros((6, 1)) #
           Initialize with proper dimensions
9
10        def update(self, robot: Manipulator):
11            DoF = robot.getDOF()
12            # Update Jacobean matrix - task Jacobian
13            self.J = robot.getLinkJacobian(self.link_index)[[0,1,5]].reshape
(self.task_dim, DoF)
14            # Update task error
15            self.err = self.getDesired() - robot.getJointConfiguration2D(self
.link_index)
16            # Compute feed-forward velocity
17            self.ffVel = (newDesired - self.getDesired()) / dt
18            # Set new desired
19            self.setDesired(newDesired)
20            return True

```

### 1.3 Implementation of the recursive formulation of the Task-Priority algorithm

#### Listing 5: Simulation Loop

```

1 # Simulation loop
2 def simulate(t):
3     global tasks
4     global robot
5     global PPx, PPy
6     global time, N_iter, Tt
7
8     ### Recursive Task-Priority algorithm
9     # Initialize null-space projector
10    P = np.eye(n_DoF, n_DoF)
11    # Initialize output vector (joint velocity)
12    dq = np.zeros((n_DoF, 1))
13    # Loop over tasks
14    for i in range(len(tasks)):
15        # Update task state
16        tasks[i].update(robot)
17        # Compute augmented Jacobian
18        Jbar = tasks[i].J @ P
19        # Compute task velocity
20        # Accumulate velocity

```

```

21     dq      = dq + DLS(Jbar, 0.1) @ (tasks[i].err - tasks[i].J @ dq)
22     # Update null-space projector
23     P       = P - DLS(Jbar, 0.001) @ Jbar
24     ###
25     # Update robot
26     robot.update(dq, dt)
27
28     # Update drawing
29     PP = robot.drawing()
30     line.set_data(PP[0,:], PP[1,:])
31     PPx.append(PP[0,-1])
32     PPy.append(PP[1,-1])
33     path.set_data(PPx, PPy)
34     point.set_data(tasks[0].getDesired()[0], tasks[0].getDesired()[1])
35
36     return line, path, point

```

## 1.4 Definition of different task hierarchies (lists of tasks)

Listing 6: Robot definition (3 revolute joint planar manipulator)

```

1 tasks = [
2 Position2D("End-effector position", np.array([0.0, 1.2]).reshape(2,1),
3   robot),
4 Position2D("Joint 1 position", np.array([0.0, 0.2]).reshape(2,1), robot),
5 Orientation2D("End-effector orientation", np.array(0.0), robot),
6 Configuration2D("End-effector configuration", np.array([0.0, 1.2, 0.0]).
7   reshape(3,1), robot)
8 ]

```

## 1.5 Random desired end-effector position

Listing 7: Init Function

```

1 # Simulation initialization
2 def init():
3     global tasks, N_iter
4     line.set_data([], [])
5     path.set_data([], [])
6     point.set_data([], [])
7     q1.set_data([], [])
8
9     # Set random new desired position

```

```

10     theta_rand = 2 * math.pi * np.random.rand()
11     length_rand = np.random.uniform(1.0, sum(a[0:3]))
12     tasks = [
13         Position2D("End-effector position", np.array([length_rand * np.
14             cos(theta_rand), length_rand * np.sin(theta_rand)]).reshape(2,1),
15             robot),
16         # Position2D("Joint 1 position", np.array([0.0, 0.2]).reshape
17             (2,1), robot)
18         # Orientation2D("End-effector orientation", np.array(0.0), robot)
19         # Configuration2D("End-effector configuration", np.block([np.
20             array([length_rand * np.cos(theta_rand), length_rand * np.sin(
21             theta_rand)]), 0.0]).reshape(3,1), robot),
22     ]
23     # Set number of iteration
24     N_iter += 1
25
26     return line, path, point, q1

```

## 2 Exercise 02

The goal of this exercise is to extend the code of Exercise 1, adding new features that allow for more flexible task definition. These features include: link selection for position and orientation tasks, gain matrices (with associated weighted DLS implementation) and the feed-forward velocity component (tracking).

Implementation on python code:

### 2.1 Manipulator class

Listing 8: Get the transformation for a selected link

```

1  '''
2      Method that return the transformation of a selected link
3
4      Argument:
5      link_index (integer):
6
7      Returns:
8      (np.array((4,4))): transformation matrix of the selected link
9  '''
10 def getLinkTranform(self, link_index: int):
11     return self.T[link_index]

```

Listing 9: Get the Jacobian for a selected link

```

1  '''
2      Method that return the Jacobean for a selected link
3
4      Argument:
5      link_index (integer):
6
7      Returns:
8      (np.array((6,DoF))): Jacobean matrix of the selected link
9  '''
10 def getLinkJacobian(self, link_index):
11     # Transformation matrix from base to each link, until get the
12     # selected link
13     TT = []
14     for i in range(link_index+1):
15         TT.append(self.getLinkTranform(i))
16     # return Jacobean matrix for selected link
17     return jacobian(TT, self.revolute)

```

## 2.2 Task class

Listing 10: Methods work with the Gain Matrix and Feed-Forward Velocity

```

1  '''
2      Method returning the gain matrix (K).
3  '''
4  def getGainMatrix(self):
5      return self.K
6  '''
7      Method setting the gain matrix (K).
8  '''
9  def setGainMatrix(self, K):
10     self.K = K
11     return True
12
13  '''
14      Method returning the feed-forward velocity.
15  '''
16  def getFFVelocity(self):
17     return self.ffVel
18  '''
19      Method setting the gain matrix (K).
20  '''
21  def setFFVelocity(self, ffVel):

```

```

22     self.ffVel = ffVel
23     return True

```

## 2.3 Task class with task index

### Listing 11: Init Function

```

1  '''
2      Base class representing an abstract Task.
3  '''
4  class Task:
5      '''
6          Constructor.
7
8          Arguments:
9          name (string): title of the task
10         desired (Numpy array): desired sigma (goal)
11     '''
12     def __init__(self, name, desired, robot: Manipulator):
13         self.name = name # task title
14         self.sigma_d = desired # desired sigma
15         self.task_dim = np.shape(desired)[0] # Task dimension
16         self.J = np.zeros((self.task_dim, robot.getDOF()))
17         self.err = np.zeros((self.task_dim, 1))
18         # Get joint number of task from name of the task
19         self.link_index = self.name_to_link_index(name, robot.getDOF())
20
21         self.K = np.eye(self.task_dim, self.task_dim) * 1
22         self.ffVel = np.eye(self.task_dim, 1)
23     '''
24     Method getting link index.
25
26     Arguments:
27     name: string
28     DoF
29
30     Return:
31     link_index: integer
32     '''
33     def name_to_link_index(self, name: str, DoF):
34         if name.split()[0] == "End-effector":
35             # End-effector task
36             return DoF
37         else:

```



```

38     # Joint i position task
39     return int(name.split()[1])

```

## 2.4 Desired trajectory

Listing 12: Desired Trajectory

```

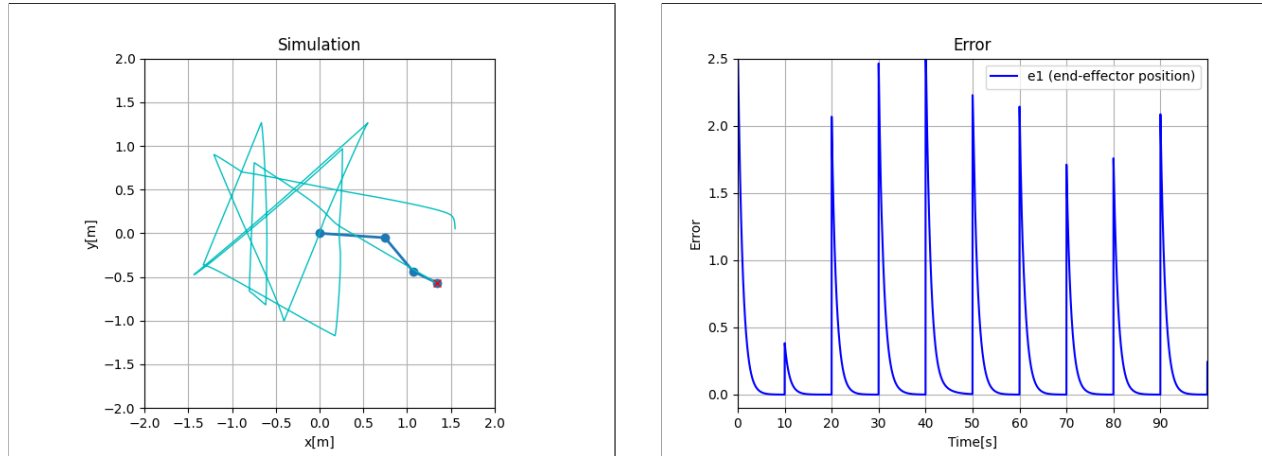
1     r_origin    = 1.0
2     r_A        = 0.0
3     r_freq     = 5.0
4     pos_ref    = 0.5
5     r          = r_origin + r_A * np.sin(r_freq*t)
6     newDesired = [(np.array([np.cos(pos_ref*t) * r, np.sin(pos_ref*t) * r
    ]).reshape(2,1)), np.array([0.0])]

```

## Result

### 3 Exercise 1:

#### 3.1 One task -> 1: end-effector position

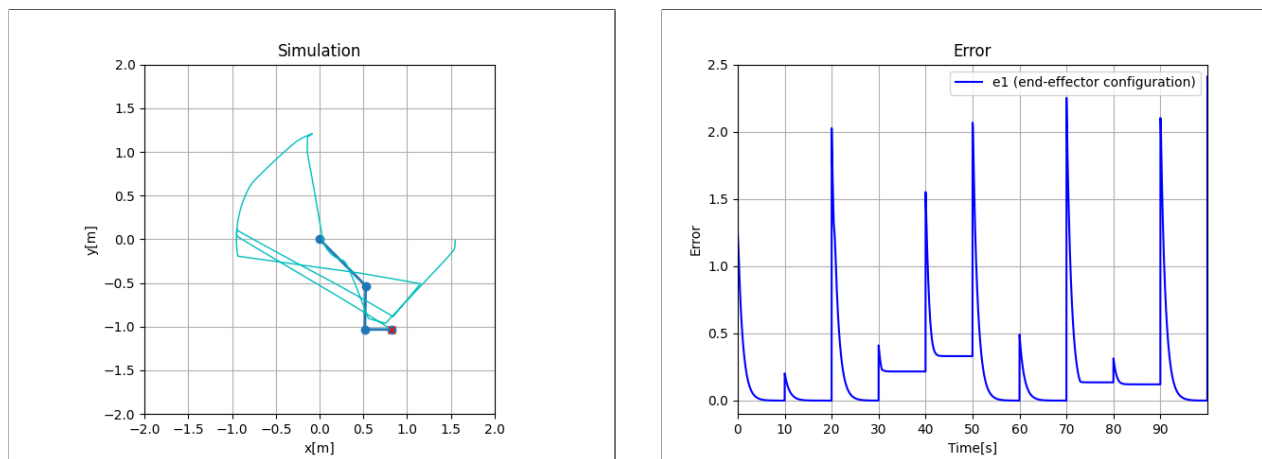


(a) Simulation of the manipulator

(b) Evolution of the TP control errors

Figure 2: One task -> 1: end-effector position

#### 3.2 One task -> 1: end-effector configuration

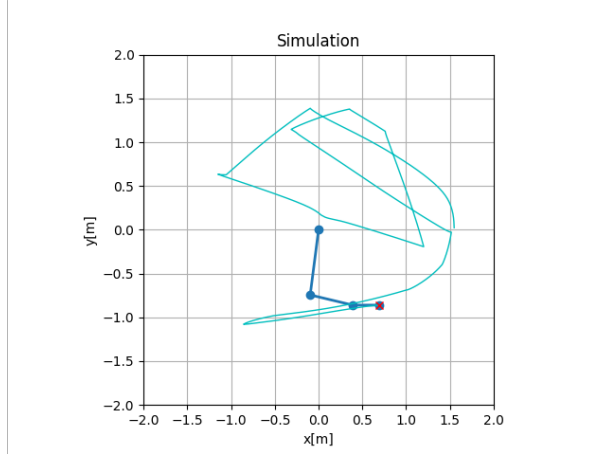


(a) Simulation of the manipulator

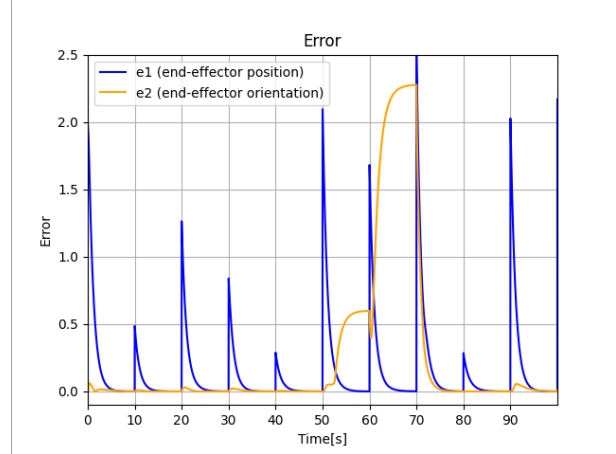
(b) Evolution of the TP control errors

Figure 3: One task -> 1: end-effector configuration (Position: Random; Heading: 0.0 deg)

### 3.3 Two tasks -> 1: end-effector position, 2: end-effector orientation



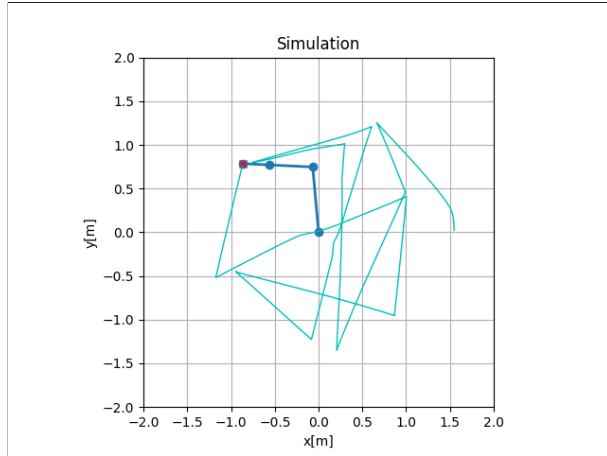
(a) Simulation of the manipulator



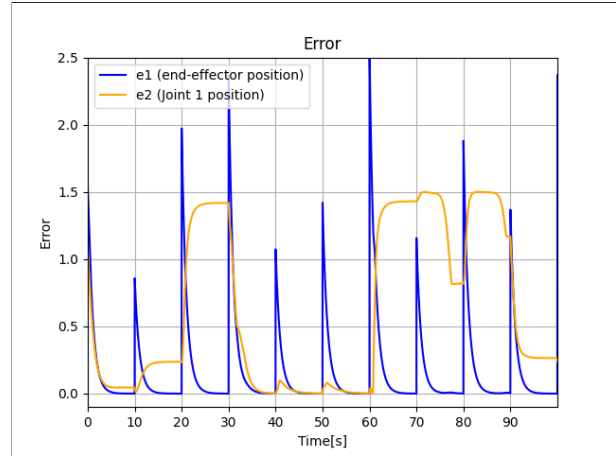
(b) Evolution of the TP control errors

Figure 4: Two tasks -> 1: end-effector position, 2: end-effector orientation (Position: Random; Heading: 0.0 deg)

### 3.4 Two tasks -> 1: end-effector position, 2: joint 1 position



(a) Simulation of the manipulator

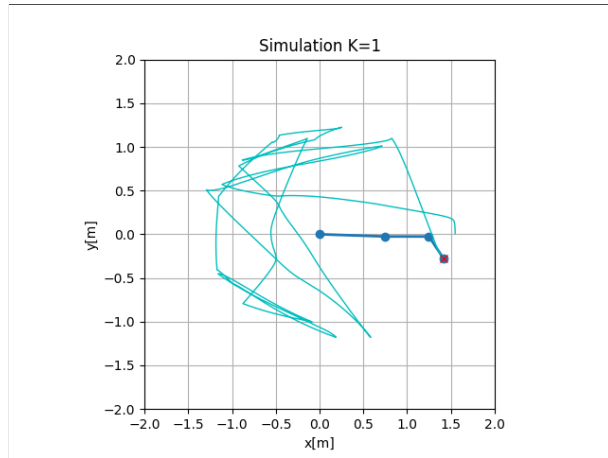


(b) Evolution of the TP control errors

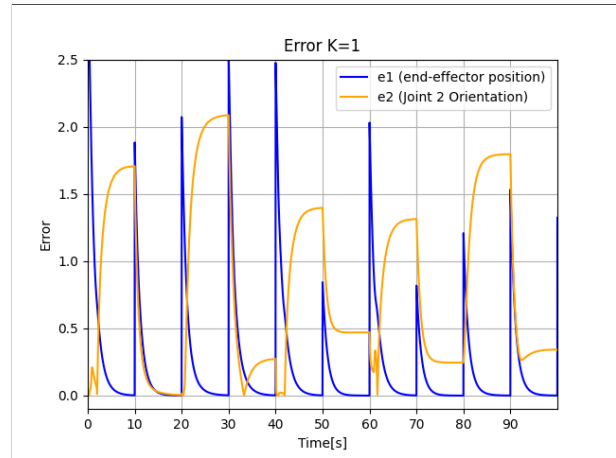
Figure 5: Two tasks -> 1: end-effector position, 2: joint 1 position (End-Effector Position: Random; Joint 1 Position: [0.0, 0.75])

## 4 Exercise 2: Two tasks -> 1: end-effector position, 2: joint 2 orientation

### 4.1 $K = 1.0$



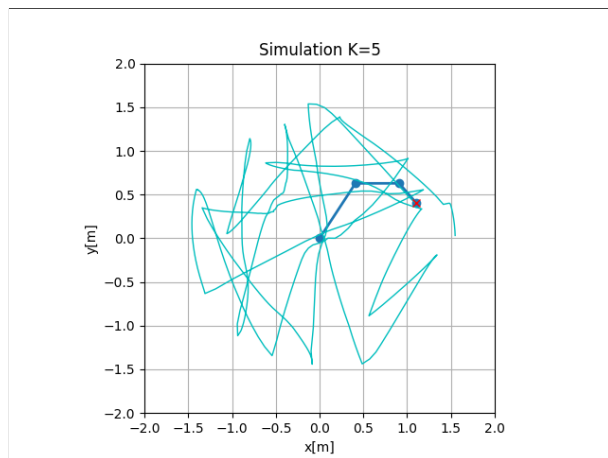
(a) Simulation of the manipulator



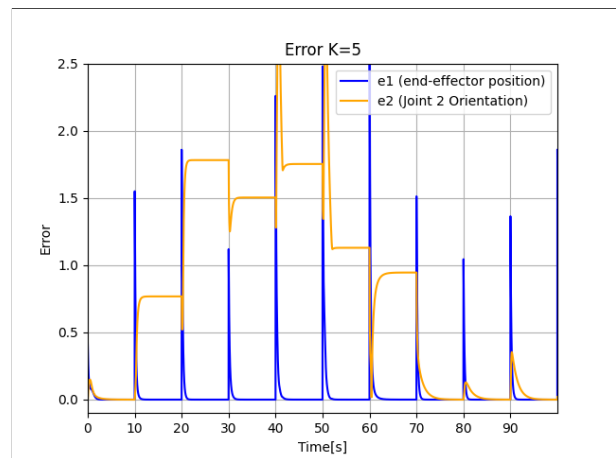
(b) Evolution of the TP control errors

Figure 6:  $K=1.0$

### 4.2 $K = 5.0$



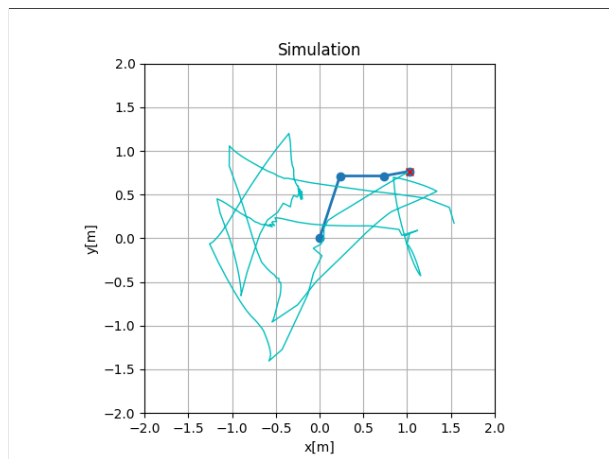
(a) Simulation of the manipulator



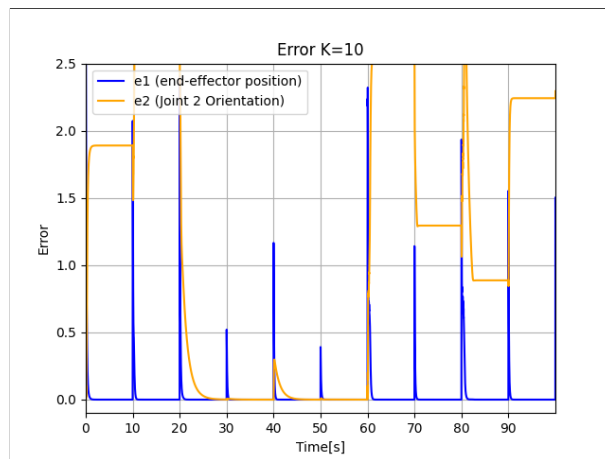
(b) Evolution of the TP control errors

Figure 7:  $K=5.0$

### 4.3 $K = 10.0$



(a) Simulation of the manipulator



(b) Evolution of the TP control errors

Figure 8:  $K=10.0$

#### 4.4 Tracking

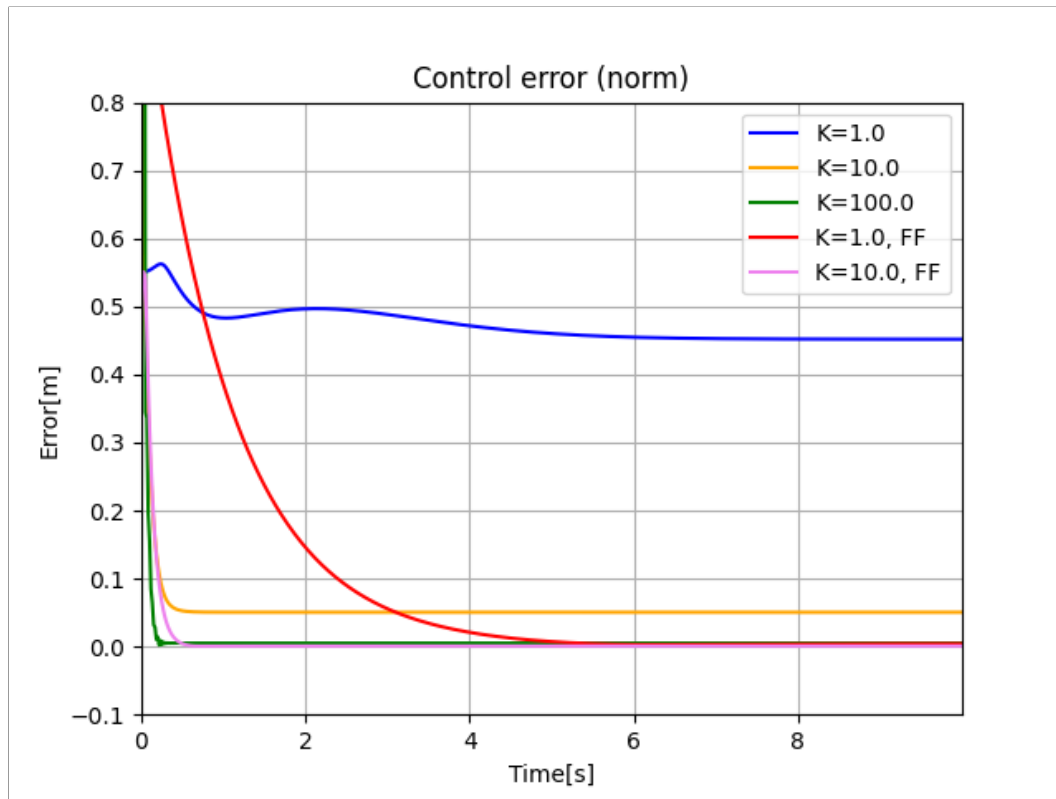


Figure 9: Error analysis: Evolution of the TP control errors

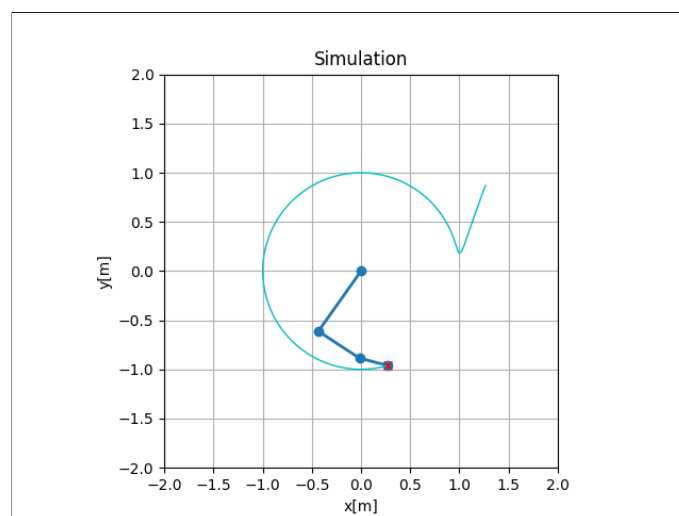


Figure 10: Simulation of the manipulator