

# Optimization and Learning for Robot Control: Assignment 2

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# 1 Path Tracking

In this first experiment, we see how the simplest form of path tracking performs, as a baseline.

## 1.1 Numerical results

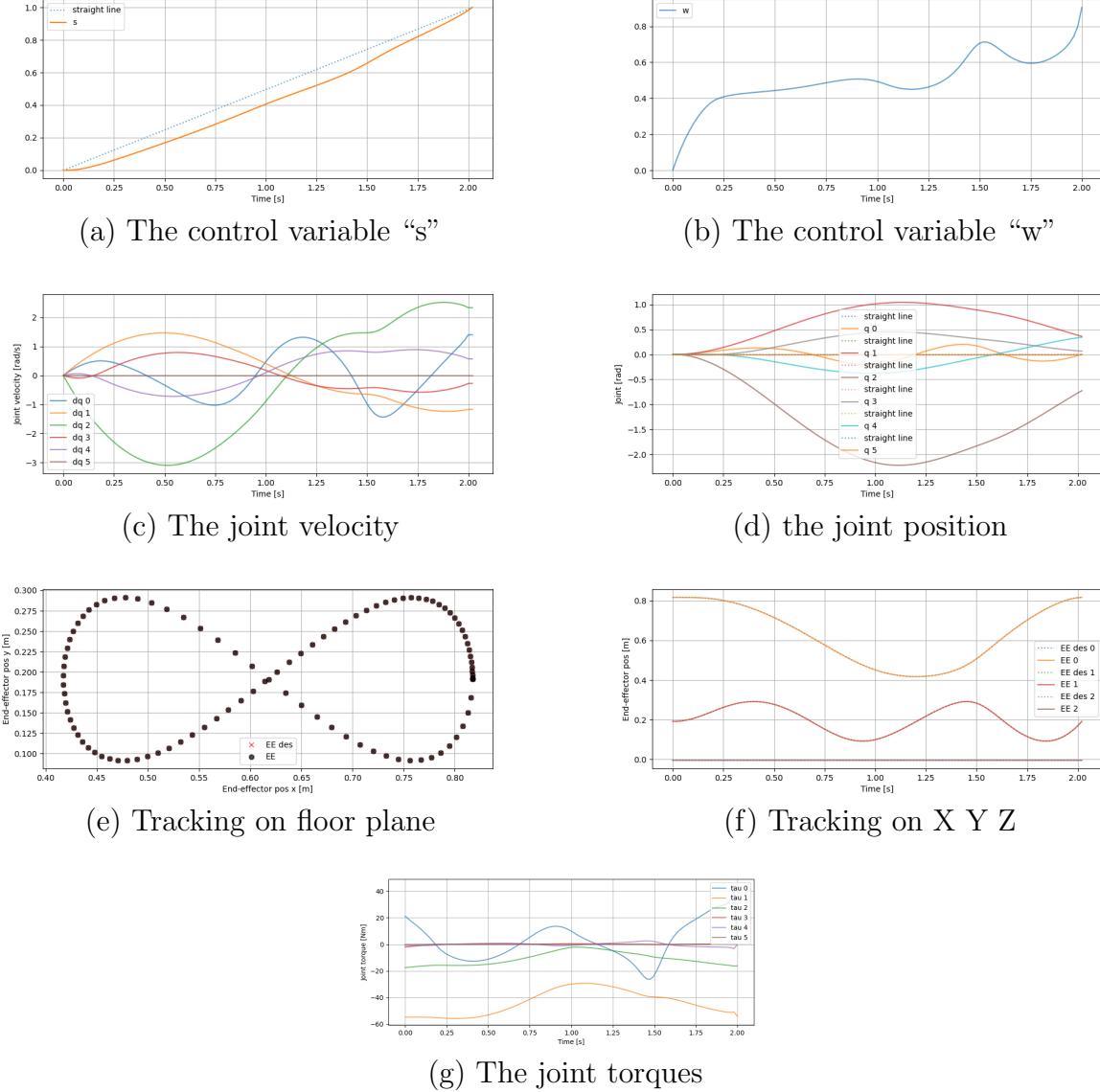


Figure 1: experiment result of “Path Tracking”

## 1.2 Observation on results

We can see in sub-figures e and f that the tracking is perfect (as we would expect from path tracking).

From the sub-figures a and b, we can instead note that the optimizer has decided to go slower (aka having a lower  $w$  value) at the beginning and at the end of the movement. These seem to be the most critical zones to guarantee good tracking.

## 2 Path Tracking + Cyclical Constraint

In this experiment, we tried to make the robot perform a cyclical task by adding a penalty that minimizes the difference between the initial state and the final state of the robot.

### 2.1 Numerical results

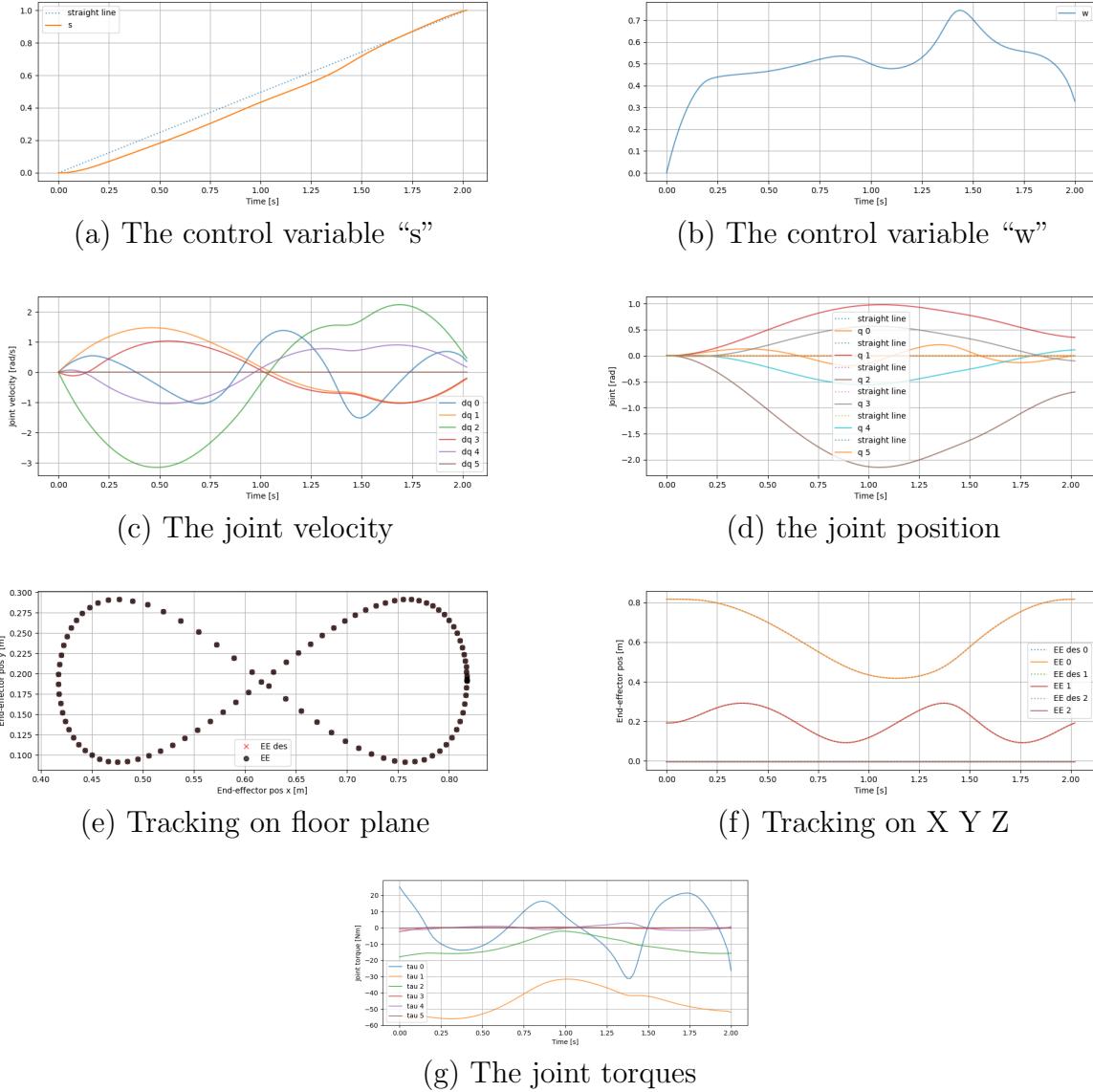


Figure 2: experiment result of “Path Tracking + Cyclical Constraint”

### 2.2 Observation on results

Here the results are mixed. The joint velocities are decent; in fact, we can see how they approach zero by the end of the simulation. However, the joint positions are not much different from those of the previous experiment, indicating that the solver is not really able to end in the same position it started, while also keeping perfect tracking (which is required, given in path tracking we use equality constraints).

### 3 Trajectory Tracking + Cyclical Constraint

In this section, we applied trajectory tracking instead of path tracking to observe the difference with the previous experiment. We still maintained the so-called “cyclical constraint”.

#### 3.1 Numerical results

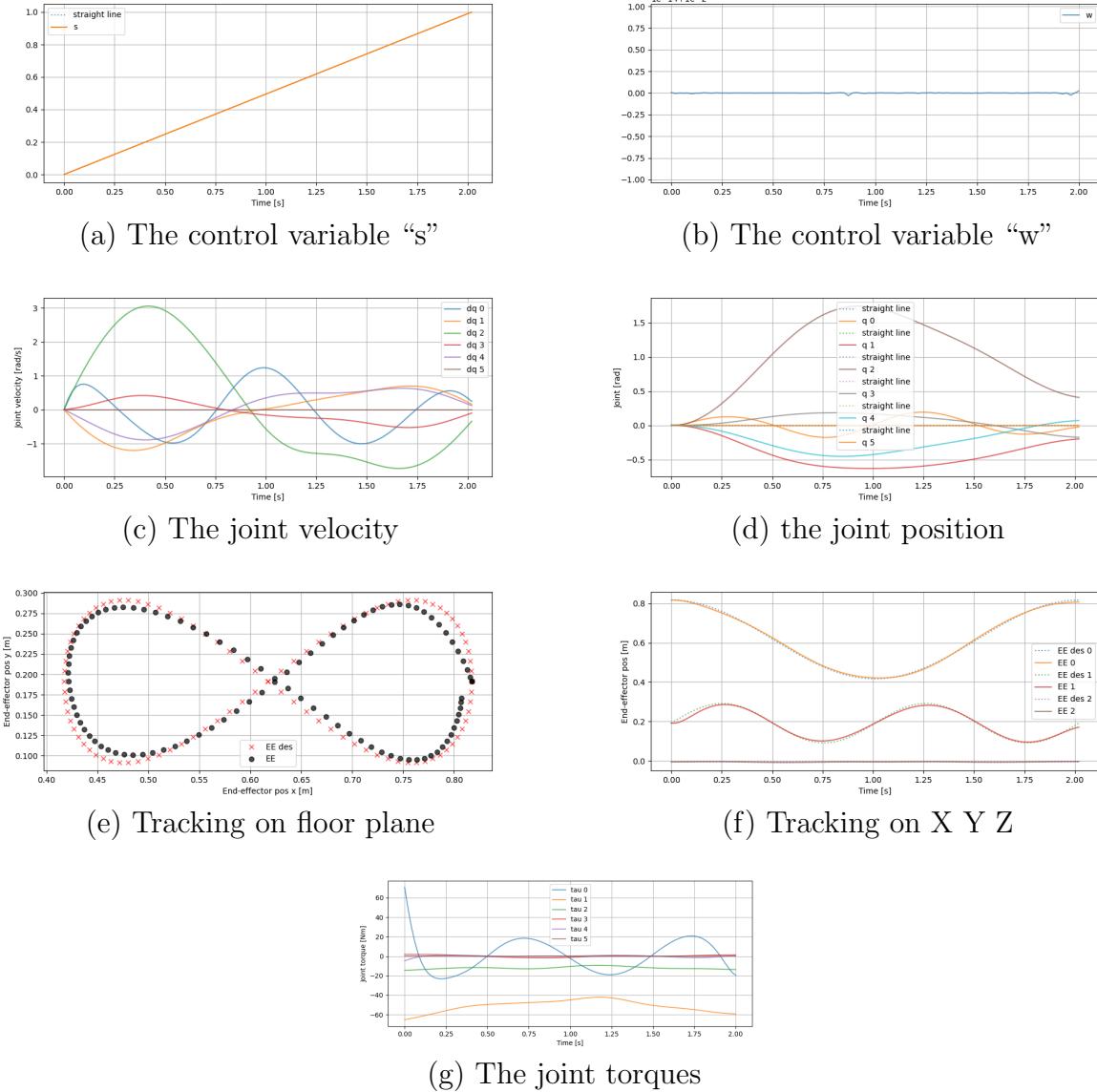


Figure 3: experiment result of “Trajectory Tracking + Cyclical Constraint”

#### 3.2 Observation on results

Here we can see in sub-figure **e** how the tracking loses a bit of accuracy. However, we can also see in the sub-figures **c** and **d** how both joint velocity and joint position revert back to the original state as the simulation approaches the end. This is because by choosing trajectory tracking, we are allowing the robot to make smaller deviations from its original path, and this makes it possible to optimize other tasks (such as the “cyclical” one).

To achieve this, we had to change the “equality” constraints on the path and make them a “cost”. If we had left them as an equality while removing the flexibility of selecting  $\mathbf{w}$ , the optimizer would almost certainly fail (unless the tracking task was trivial).

# 4 Path Tracking + Cyclical Constraint + Min Time Formulation

In this experiment, we introduced a min-time Formulation constraint to the previous path tracking task (by letting the solver optimize the time step).

## 4.1 Numerical results

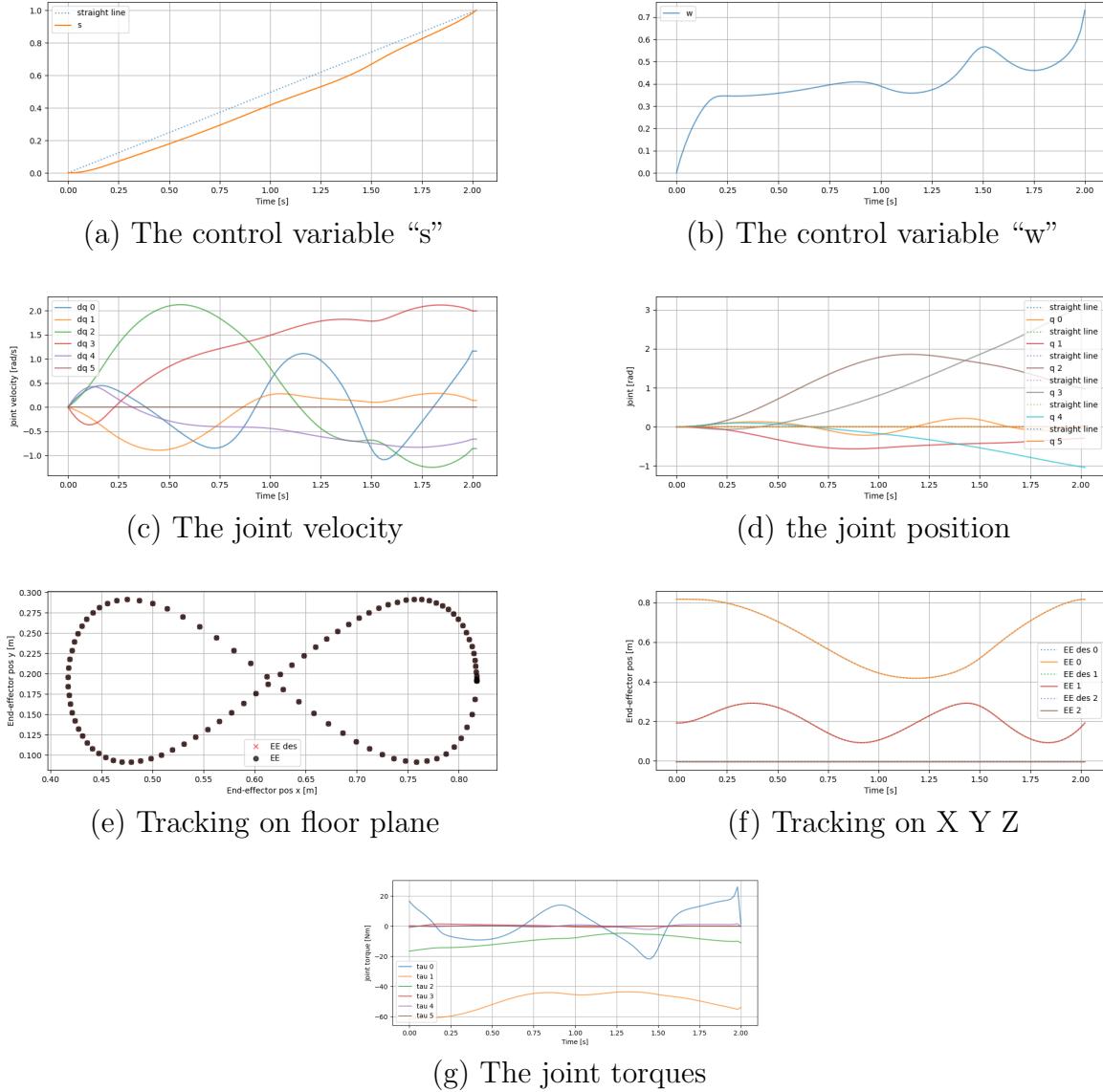


Figure 4: experiment result of “Path Tracking + Cyclical Constrain + Min Time Formulation”

## 4.2 Observation on results

We can see how the results barely change from the previous path-tracking Formulation. And the reason is easily understood by printing the time-step value after the optimization process.

We noted that the solver was actually taking more time than the previous Formulation (using a time step of 24ms instead of 20).

The simple solution to this issue was to increase the weight of the running cost (see next section).

# 5 Path Tracking + Cyclical Constraint + Min Time Formulation + Tuned Weights

This experiment has the same setup of the previous section, but the running cost's weight was set to 1 instead of 0.1.

## 5.1 Numerical results

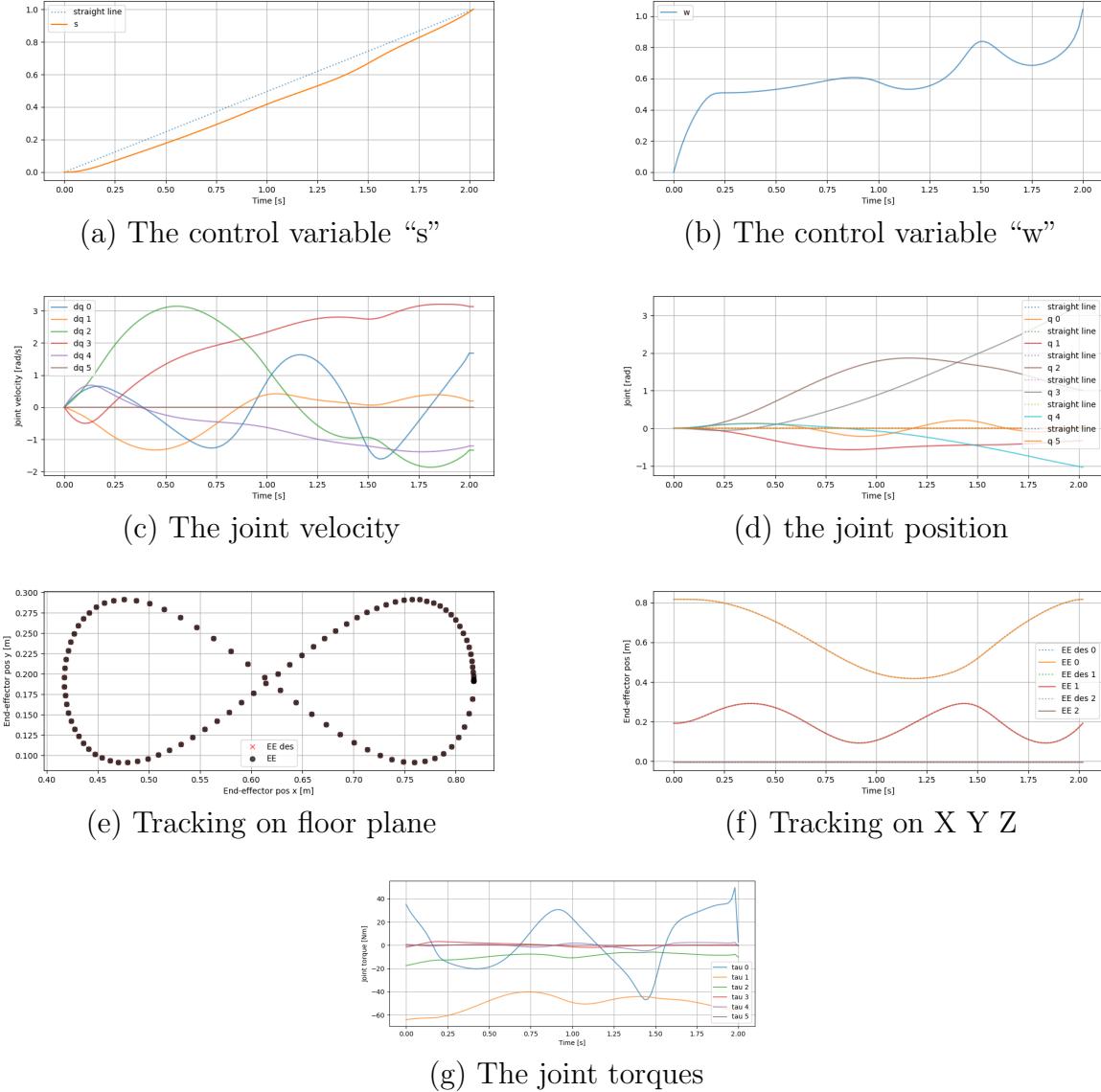


Figure 5: experiment result of “Path Tracking + Cyclical Constraint + Min Time Formulation + Tuned Weights”

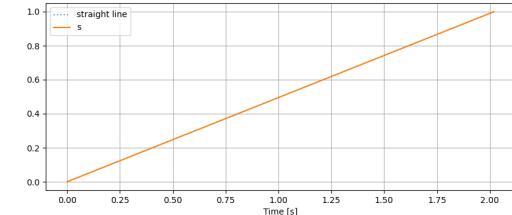
## 5.2 Observation on results

Here the solver managed to complete the trajectory in a much quicker time, as the time-step was optimized to be 14ms. Here the charts look similar to the previous ones, with the only notable difference being that the torques, accelerations, and velocities are all slightly higher, indicating that the controller has decided to get a higher penalty in the velocities, accelerations, and control inputs in order to minimize the penalty on the time-step.

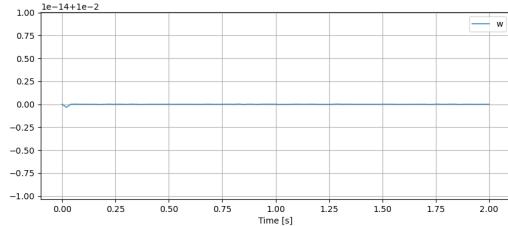
# 6 Trajectory Tracking + Cyclical Constraint + Min Time Formulation

In this experiment, we apply the same min-time Formulation that we made in the previous section to trajectory Tracking.

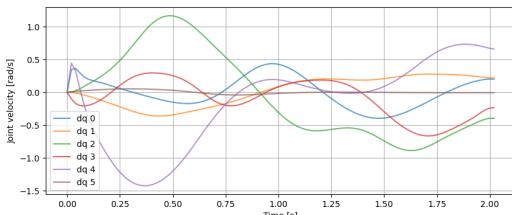
## 6.1 Numerical results



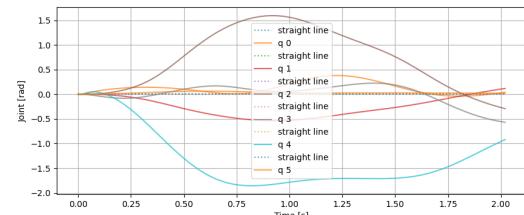
(a) The control variable “ $s$ ”



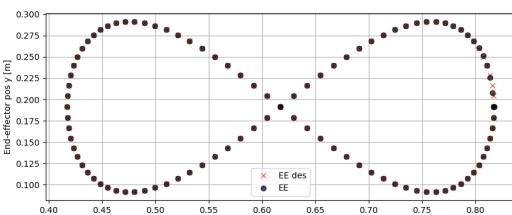
(b) The control variable “ $w$ ”



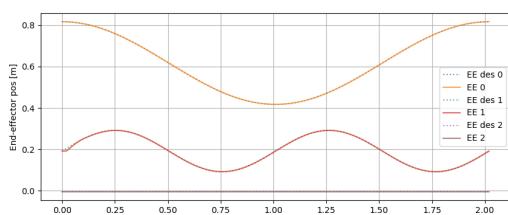
(c) The joint velocity



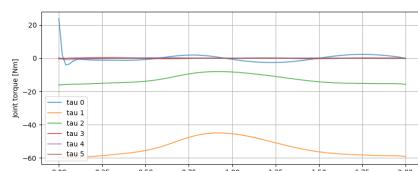
(d) the joint position



(e) Tracking on floor plane



(f) Tracking on X Y Z



(g) The joint torques

Figure 6: experiment result of “Trajectory Tracking + Cyclical Constraint + Min Time Formulation”

## 6.2 Observation on results

These are by far the best results we got. We can see how the tracking is almost perfect, even though we are using trajectory tracking instead of path tracking. The total time required to make the maneuver is also quite low (with a time-step that was around 6ms).

However, we can see from sub-plot **c** and **d** that the robot does not return to the original state quite as well as our previous best result did.

Joint velocity and accelerations are higher than previous tests, but not extremely higher. This may sound absurd at first, however, we can make sense of it by looking at the joint position sub-plot **d**. We see how this chart looks a bit different from the equivalent charts in previous experiments, and this implies that the robot is using a different posture to achieve the task. And we can assume that this posture is better for a quick execution.

Why the solver did not use this posture for previous optimization tasks is hard to say, as solvers like the one we used often function as a “black box”. However, we can hypothesize that this configuration was not used in path tracking because it made it impossible to track perfectly, and it also was not used during the previous trajectory tracking test, because the “cyclical constraint” had a higher weight relative to the others, and maybe this configuration was not chosen because it provided a worse end position.