

Robotics

Exercise 3

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1 Motion profiles (2 points)

Construct a motion profile that accelerates constantly in the first quarter of the trajectory, then moves with constant velocity, then decelerates constantly in the last quarter. Write down the equation $MP(s) : [0, 1] \mapsto [0, 1]$.

2 Verify some things from the lecture (6 points)

a) On slide 02:30 we derived the basic inverse kinematics law. Verify that (assuming linearity of ϕ , i.e., $J\delta q = \delta y$) for $C \rightarrow \infty$ the desired task is fulfilled exactly, i.e., $\phi(q^*) = y^*$. By writing $C \rightarrow \infty$ we mean that C is a matrix of the form $C = \varrho C_0$, $\varrho \in \mathbb{R}$, and we take the limit $\varrho \rightarrow \infty$.

b) On slide 02:36 there is a term $(\mathbf{I} - J^\# J)$ called nullspace projection. Verify that for $\varrho \rightarrow \infty$ (and $C = \varrho \mathbf{I}$) and any choice of $\delta a \in \mathbb{R}^n$

$$\delta q = (\mathbf{I} - J^\# J)\delta a \Rightarrow \delta y = 0$$

(assuming linearity of ϕ , i.e., $J\delta q = \delta y$). Note: this means that any choice of δa , the motion $(\mathbf{I} - J^\# J)\delta a$ will not change the “endeffector position” y . (2 P)

c) On slide 02:48 it says that

$$\begin{aligned} \operatorname{argmin}_q \|q - q_0\|_W^2 + \|\Phi(q)\|^2 \\ \approx q_0 - (J^\top J + W)^{-1} J^\top \Phi(q_0) = q_0 - J^\# \Phi(q_0) \end{aligned}$$

where the approximation \approx means that we use the local linearization $\Phi(q) = \Phi(q_0) + J(q - q_0)$. Verify this. (2 P)

3 IK in the simulator (6 points)

Installation instructions:

1. On github <https://github.com/MarcToussaint/robotics-course> you can find the course repository and an instruction on how to install it.
2. To make sure you have an updated version of the repository, run `'git pull'` and `'git submodule update'`
3. For python you can run: `'jupyter-notebook py/01-kinematics/01-kinematics.ipynb'`
4. For C++ run: `'cd cpp/01-kinematics', 'make', './x.exe -mode 2'`

The goal of this task is to reach the coordinates $y^* = (-0.2, -0.4, 1.1)$ with the right hand of the robot. Assume $W = \mathbf{I}$ and $\sigma = .01$.

- a) The provided code already generates a motion using inverse kinematics $\delta q = J^\# \delta y$ with $J^\# = (J^\top J + \sigma^2 W)^{-1} J^\top$. Record the task error, that is, the deviation of hand position from y^* after each step. You can plot the error using `'plt.plot(err)'` and `'plt.show()'` in python or `'gnuplot(err)'` in C++ (err is the array of errors). Why is it initially large? (1 P)
- b) Try to do 100 smaller steps $\delta q = \alpha J^\# \delta y$ with $\alpha = .1$ (each step starting with the outcome of the previous step). How does the task error evolve over time? (1 P)
- c) Generate a nice trajectory composed of $T = 100$ time steps. Interpolate the target linearly $\hat{y} \leftarrow y_0 + (t/T)(y^* - y_0)$ in each time step. How does the task error evolve over time? (2 P)
- d) Generate a trajectory that moves the right hand in a circle centered at $(-0.2, -0.4, 1.1)$, aligned with the xz -plane, with radius 0.2. (2 P)