

# Motion Planner – Python library overview

for any questions, please contact [stephen-monnet@hotmail.com](mailto:stephen-monnet@hotmail.com)

## 1. Main goals

This library is based on the [mpc\\_motion\\_planner](#) project (C++), which aims to generate a trajectory  $x(\cdot) = \{q(\cdot), \dot{q}(\cdot), \ddot{q}(\cdot)\}$  between two arbitrary points  $x_0 = \{q_0, \dot{q}_0, \ddot{q}_0\}$  and  $x_f = \{q_f, \dot{q}_f, \ddot{q}_f\}$  for the Franka Panda robot with 7 degrees of freedom. This trajectory can be generated either (a) using [ruckig](#) or (b) using [polympc](#) with ruckig as the starting solution. It should be noted that (a) only allows for considering box constraints on velocity, acceleration, and jerk, while (b) allows for adding non-linear constraints such as the minimum height of the tool or the maximum permissible torque.

In addition to making the functionalities of mpc\_motion\_planner accessible in Python, this library extends its capabilities to the Kuka iiwa7 and Kuka iiwa14 robots, offering numerous possibilities for manipulating and analyzing the generated data.

## 2. Model

### a. Dynamic

For polympc, the robot's dynamics are simply modelled as a double integrator between the input  $u$  and the joint positions  $q$  (thus,  $u = \ddot{q}$ ):

$$\dot{x} = \begin{pmatrix} \dot{q} \\ \ddot{q} \end{pmatrix} = \begin{pmatrix} \mathbf{0}_7 & \mathbf{I}_7 \\ \mathbf{0}_7 & \mathbf{0}_7 \end{pmatrix} \cdot \begin{pmatrix} q \\ \dot{q} \end{pmatrix} + \begin{pmatrix} \mathbf{0}_7 \\ \mathbf{I}_7 \end{pmatrix} \cdot u$$

### b. Objective Function (Polympc)

The objective of the NOCP (Nonlinear Optimal Control Problem) is to minimize the total duration of the trajectory that connects the initial state  $x_0$  to the final state  $x_f$ . For more information on the formulation, refer to [Jennings problem](#).

### c. Ruckig constraints :

$\dot{q}_{min} \leq \dot{q}(\cdot) \leq \dot{q}_{max}$	:	Joint Velocity
$\ddot{q}_{min} \leq \ddot{q}(\cdot) \leq \ddot{q}_{max}$	:	Joint Acceleration
$\dddot{q}_{min} \leq \dddot{q}(\cdot) \leq \dddot{q}_{max}$	:	Joint Jerk

### d. Polympc constraints :

$q_{min} \leq q(\cdot) \leq q_{max}$	:	Joint Position
$\dot{q}_{min} \leq \dot{q}(\cdot) \leq \dot{q}_{max}$	:	Joint Velocity
$\ddot{q}_{min} \leq \ddot{q}(\cdot) \leq \ddot{q}_{max}$	:	Joint Acceleration
$\tau_{min} \leq \tau(\cdot) \leq \tau_{max}$	:	Joint Torque
$x_{ee}(\cdot) = [0 \ 0 \ 1] \cdot FK_{xyz}(q(\cdot)) \geq h_{min}$	:	End Effector Height

### 3. Main Architecture of the Framework

The main task handled by the `mpc_motion_planner` block is to set up the Nonlinear Optimal Control Problem (NOCP) in a form that `polympc` can solve. It is responsible for evaluating the state of constraints and defining the dynamics of the system  $\dot{x} = f(x, u)$ .

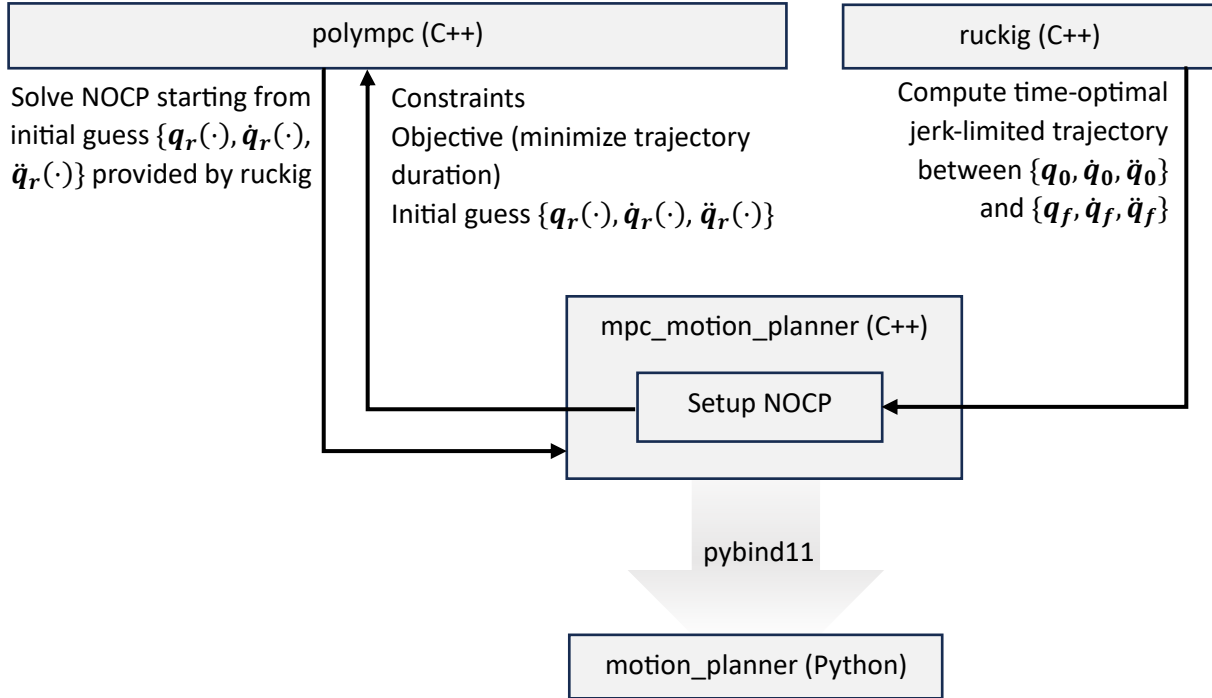


Figure 1 : Framework Architecture

### 4. Constraints Margin

For practical reasons, it is important to incorporate a safety margin on certain constraints. For example, for a state constraint  $s_{min} \leq s \leq s_{max}$ , new bounds will be calculated as  $s_{min} \leq s'_{min} \leq s \leq s'_{max} \leq s_{max}$ :

$$s'_{min} = \frac{s_{min} + s_{max}}{2} - \frac{(s_{max} - s_{min})}{2} \cdot \eta = s_{min} + \frac{(1 - \eta) \cdot (s_{max} - s_{min})}{2}$$

$$s'_{max} = \frac{s_{min} + s_{max}}{2} + \frac{(s_{max} - s_{min})}{2} \cdot \eta = s_{max} - \frac{(1 - \eta) \cdot (s_{max} - s_{min})}{2}$$

Note that if  $s_{max} = -s_{min}$ , then :

$$s'_{min} = s_{min} \cdot \eta$$

$$s'_{max} = s_{max} \cdot \eta$$

In the case of symmetrical constraints such as velocity, acceleration, or torque constraints. Please mind the notation, which might be a bit confusing, where  $\eta$  is called “margin” while it is usually defined in the literature as  $(1 - \eta)$ .

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## 5. Classes

The Python library is organized with two main classes:

### a. Trajectory

A trajectory object stores mainly 5 vector values (time  $t$ , joint position  $\mathbf{q}$ , joint velocity  $\dot{\mathbf{q}}$ , joint acceleration  $\ddot{\mathbf{q}}$ , joint torque  $\boldsymbol{\tau}$ ) and provides convenient methods to assess constraint satisfaction.

### b. MotionPlanner

A MotionPlanner object is used to interface with the C++ class that corresponds to the desired robot (Panda, Kuka-iiwa7, Kuka-iiwa14). It provides methods to set the constraint limits and boundary conditions ( $\{\mathbf{q}_0, \dot{\mathbf{q}}_0, \ddot{\mathbf{q}}_0\}$  and  $\{\mathbf{q}_f, \dot{\mathbf{q}}_f, \ddot{\mathbf{q}}_f\}$ ), solve a trajectory using ruckig and/or polympc and compute inverse / forward kinematics.

For more information about how to use these objects, please refer to the GitHub repository : [mpc\\_motion\\_planner/pyMPC/howToUse.ipynb](https://github.com/mpc-motion-planner/pyMPC/howToUse.ipynb)