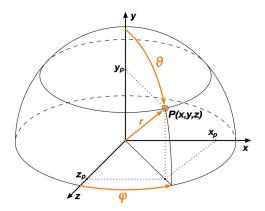
## Exercise: Optimal Control of a Robot Arm

This exercise aims at steering the robot arm to perform time-optimal point-to-point motions.





We model the robot arm in 3D as follows

$$\ddot{\varphi}(t) = \frac{u_{\varphi}(t)}{r \cdot m} - \underbrace{2\cot(\vartheta(t))\dot{\varphi}(t)\dot{\vartheta}(t)}_{\text{due to coriolis forces}}$$
 
$$\ddot{\vartheta}(t) = \underbrace{\sin(\vartheta(t))g}_{\text{due to gravitation}} + \underbrace{\frac{u_{\vartheta}(t)}{r \cdot m}}_{\text{due to centrifugal forces}} + \underbrace{\sin(\vartheta(t))\cos(\vartheta(t))\dot{\varphi}(t)^2}_{\text{due to centrifugal forces}}$$

with gravitational constant  $g = 9.81 \,[\text{m/s}^2]$  and mass  $m = 1 \,[\text{kg}]$  and arm length  $r = 1 \,[\text{m}]$ . Then, we can summarize our dynamic system as

$$\dot{x}(t) = f(x(t), u(t)) \quad \text{with } \begin{cases} x(t) = (\varphi(t), \dot{\varphi}(t), \vartheta(t), \vartheta(t)) \in \mathbb{R}^4 \\ u(t) = (u_{\varphi}(t), u_{\vartheta}(t)) \in \mathbb{R}^2 \end{cases}$$

where the control inputs are defined by the motor torques  $u_{\varphi}(t)$  and  $u_{\vartheta}(t)$  in  $[N \cdot m]$ . The optimal control problem is formulated in continuous time below

$$\begin{array}{lll} \text{minimize} & T & \text{Objective} \\ \text{subject to} & \forall t \in [0,T], \quad \dot{x}(t) = f(x(t),u(t)) & \text{Nonlinear dynamics} \\ & x(0) = \begin{pmatrix} 0 \\ 0 \\ \frac{\pi}{2} \\ 0 \end{pmatrix}, \ x(T) = \begin{pmatrix} \frac{\pi}{2} \\ 0 \\ \frac{\pi}{2} \\ 0 \end{pmatrix} & \text{Boundary constraints} \\ & -10 \left[\mathbf{N} \cdot \mathbf{m}\right] \leq u_{\varphi}(t) \leq 10 \left[\mathbf{N} \cdot \mathbf{m}\right] & \text{Input} \\ & -20 \left[\mathbf{N} \cdot \mathbf{m}\right] \leq u_{\vartheta}(t) \leq 20 \left[\mathbf{N} \cdot \mathbf{m}\right] & \text{Constraints} \end{array}$$

## Implementation tasks:

- 1. consider piece-wise constant control inputs and then, discretize the continuous-time dynamic by implementing a RK4 integrator;
- 2. consider the input constraints at each sampling point and then, formulate the OCP as an NLP by using Casadi.
- 3. Solve the resulting NLP by using IPOPT interfaced in Casadi, and visualize the results.