# ROB315 Modélisation et Commande des Robots Manipulateurs

TP1 - Direct and inverse kinematics

TP2 - Dynamics and control



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# 1 Introduction

We propose to study the geometric and kinematic modeling of a manipulator arm developed by the Interactive Robotics Laboratory of the CEA List (Figure 1.1). This robot, which kinematic chain is of serial type, has 6 revolute joints  $(j_i \text{ with } i = 1,...,6)$ .

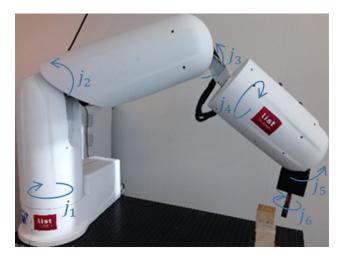


FIGURE 1.1 – Manipulator arm developed by the Interactive Robotics Laboratory of the CEA List

The numerical values of the robot parameters are specified in the table below.

Parameters	Numerical values	Type of parameter
$d_3$	$0.7\mathrm{m}$	Geometric parameter
$r_1$	0.5m	Geometric parameter
$r_4$	$0.2 \mathrm{m}$	Geometric parameter
$r_E$	$0.1 \mathrm{m}$	Geometric parameter

Table 1 – Robot parameters

The main  $MatLab^{TM}$  functions used to design the robotic arm in this project can be visualized in Appendix A.

# 2 Direct geometric model

# 2.1 Q1. Robot's geometry

First the geometry of the robot was established, identifying the frame of each of its joints according to the MDH convention. In the figure below, the frames attached to each link of the manipulator robot can be seen.

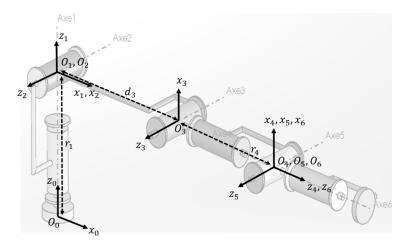


Figure 2.1 – Description of the robot's geometry

All robot joints are revolute joints, which means that the relative motion between two bodies is reduced to a pure rotation along a common axis.

# 2.2 Q2. Geometric parameters

In order to describe the working space of the manipulator arm, each of its joints has its associated geometrical parameters. These parameters can be seen in table 2.

Joint (i)	$\alpha_i$	$\mathbf{d}_i$	$ heta_i$	$\mathbf{r}_i$
1	0	0	$q_1$	$\mathbf{r}_1$
2	$\pi/2$	0	$q_2$	0
3	0	$d_3$	$q_3 + \pi/2$	0
4	$\pi/2$	0	$q_4$	$r_4$
5	$-\pi/2$	0	$q_5$	0
6	$\pi/2$	0	$q_6$	0
Е	0	0	0	$r_E$

Table 2 – Geometric parameters of the robot

The parameters d and r represent the distance of the  $j_i$  joint from the  $j_{i-1}$  one in the direction x and z respectively. Meanwhile, the parameters  $\alpha$  and  $\theta$  represent the possible angles of rotation of each joint in relation to the x and z axes respectively.

## 2.3 Q3. Direct Geometric Model (DGM) of the robot

First, in order to calculate the homogenous transformation matrix  $\overline{g}$  between two successive frames, a MatLab function TransformMatElem( $\alpha_i, d_i, \theta_i, r_i$ ) has been defined using the following formula.

$$\begin{split} \overline{g}_{(i-1)i} &= R_x, \alpha_i Trans(x,d_i) R_z, \theta_i Trans(z,r_i) \\ &= \begin{pmatrix} \cos(\theta_i) & -\sin(\theta_i) & 0 & d_i \\ \cos(\alpha_i) \sin(\theta_i) & \cos(\alpha_i) \cos(\theta_i) & -\sin(\alpha_i) & -r_i \sin(\alpha_i) \\ \sin(\alpha_i) \sin(\theta_i) & \sin(\alpha_i) \cos(\theta_i) & \cos(\alpha_i) & r_i \cos(\alpha_i) \\ 0 & 0 & 0 & 1 \end{pmatrix} \\ &= \begin{pmatrix} R_{(i-1)i} & p_{(i-1)i} \\ 0_{1x3} & 1 \end{pmatrix} \end{split}$$

where  $R_{(i-1)i}$  is the rotation matrix representing the frame  $R_{(i-1)} = (O_{(i-1)}x_{(i-1)}y_{(i-1)}z_{(i-1)})$  with respect to the frame  $R_i = (O_ix_iy_iz_i)$  and  $p_{(i-1)i}$  is the column vector that represents the translation from  $O_{(i-1)}$  to  $O_i$ .

Then a new function  $\mathsf{ComputeDGM}(\alpha,d,\theta,r)$  was developed to compute the direct geometry model of the system. This function aims to calculate the  $\overline{g}_{0N}$  transformation matrix through which it is possible to find the geometrical parameters of a given frame of the manipulator arm link chain. This transformation matrix is obtained by the following recursive formula.

$$\overline{g}_{0N}(q) = \overline{g}_{01}(q_1)...\overline{g}_{(i-1)i}(q_i)...\overline{g}_{(N-1)N}(q_N)$$
(1)

where q is the vector of joint variables.

Finally, it was desired to calculate the position and orientation of the robot's end-effector. Thus, in addition to the geometrical parameters of the system, the end-effector geometrical parameters (seen in the last line of table 1) were used as the input arguments of the GDM function. The function  $\mathtt{TransformMatElem}(\alpha_i, d_i, \theta_i, r_i)$  was used within function  $\mathtt{ComputeDGM}(\alpha, d, \theta, r)$  to calculate the homogenous transformation matrix of each frame.

#### 2.4 Q4. Position and orientation of end-effector frame

Now the position and orientation of the end-effector frame is to be calculated given two joint configurations  $q_i = [-\frac{\pi}{2}, 0, -\frac{\pi}{2}, -\frac{\pi}{2}, -\frac{\pi}{2}, -\frac{\pi}{2}]^t$  and  $q_f = [0, \frac{\pi}{4}, 0, \frac{\pi}{2}, \frac{\pi}{2}, 0]^t$ . To this end the function to compute the DGM, created in the previous section, is implemented.

#### 2.4.1 For configuration $q_i$

$$P = [-0.10 - 0.70 \quad 0.30]^{T}$$

$$n = [-0.58 - 0.58 \quad 0.58]^{T}$$

$$\phi = 2.09$$

#### 2.4.2 For configuration $q_f$

$$P = [0.63 - 0.10 \quad 1.13]^{T}$$

$$n = [0.28 \quad 0.68 - 0.68]^{T}$$

$$\phi = 2.59$$

# 2.5 Q5. Visualization of position and orientation of end-effector frame

Through the position and orientation results obtained for the end-effector frame in the last section, for both proposed configurations, it was possible to develop a function PlotFrame(q) to better visualize the frames  $R_0$  and  $R_E$ . The graphs generated by this new function are presented below. The dashed cyan line represents the translation from  $O_0$  to  $O_E$ .

# 2.5.1 For configuration $q_i$

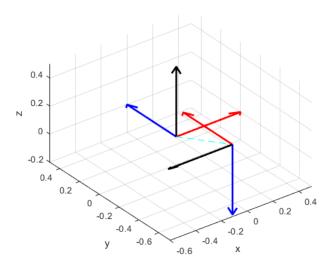


FIGURE 2.2 – Frames  $R_0$  and  $R_E$  for the configuration  $q_i$ 

# 2.5.2 For configuration $q_f$

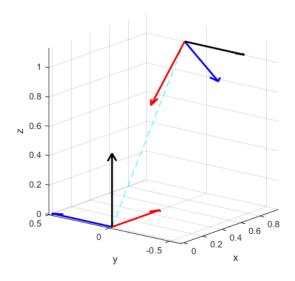


Figure 2.3 – Frames  $R_0$  and  $R_E$  for the configuration  $q_f$ 

# 3 Direct kinematic model

## 3.1 Q6. Twists calculation by the Jacobian matrix

In this question it is desired to study the direct kinematic model (DKM) of a robot, by which it is possible to determine the velocity of the end-effector  $\dot{X}$  according to the velocities of the joints  $\dot{q}$ . This is done by the following equation.

$$\dot{X} = J(q)\dot{q} \tag{2}$$

where J(q) denotes the Jacobian matrix given by  $\frac{\partial X}{\partial q}$ .

In order to calculate the linear and angular twist components of the velocity of a given frame (equation 3.1.2), the Jacobian matrix of equation 2 can be written by equation 5.

$${}^{0}\mathcal{V}_{0,E} = \begin{bmatrix} {}^{0}V_{0,E}(O_{E}) \\ {}^{0}\omega_{0,E} \end{bmatrix} = \begin{bmatrix} {}^{0}J_{v}(q) \\ {}^{0}J_{\omega}(q) \end{bmatrix} \dot{q} = {}^{0}J(q)\dot{q}$$

$$(3)$$

$${}^{0}J(q) = [{}^{0}J_{0}(q), ..., {}^{0}J_{i}(q), ..., {}^{0}J_{N}(q)]$$

$$(4)$$

where  ${}^{0}J_{i}(q)$  is the cartesian velocity due to the i<sup>th</sup> joint given in the  $\mathcal{R}_{0}$  frame.

The  ${}^{0}J_{i}(q)$  of each joint can be calculated as the following:

$${}^{0}J(q) = \begin{cases} \begin{bmatrix} R_{0i}Z_{i} \\ 0_{3x1} \end{bmatrix} & \text{if the } \mathbf{i}^{th} \text{ joint is prismatic} \\ \\ \begin{bmatrix} R_{0i}(Z_{i} \times p_{iN}) \\ R_{0i}Z_{i} \end{bmatrix} & \text{if the } \mathbf{i}^{th} \text{ joint is revolute} \end{cases}$$

As in the case under study all joints are revolute, the calculation of all the Jacobian matrices will be performed by the second method, where  $R_{0i}$  is the rotation matrix of frame  $R_i$ ,  $Z_i$  is the unit vector of the i<sup>th</sup> joint and  $p_{iN}$  is the position vector from origin of frame  $R_i$ .

Thus a function ComputeJac( $\alpha$ ,d, $\theta$ ,r) was created to calculate the Jacobian matrix  ${}^{0}J(q)$ , by which it was possible to find the twists on the end-effector frame, using the  $q_i$  and  $q_f$  of the previous sections in addition to the joint velocities  $\dot{q} = [0.5, 1.0, 0.5, 0.5, 1.0, 0.5]$ .

#### 3.1.1 For configuration $q_i$

$$^{0}\mathcal{V}_{0,E} = [0.35, -0.10, 0.60, -0.00, -1.00, 0.00]^{T}$$

#### 3.1.2 For configuration $q_f$

$${}^{0}\mathcal{V}_{0.E} = [-0.55, \ 0.32, \ 0.46, \ 1.06, \ 0.00, \ 0.15]^{T}$$

# 3.2 Q7. Transmission of velocities

In this question the study of the transmission of velocities between the joint and the task space is proposed, only taking into account the linear velocity component  ${}^{0}J_{v}(q)$ . Therefore, the decomposition in singular values is used to calculate the velocities by the following equation:

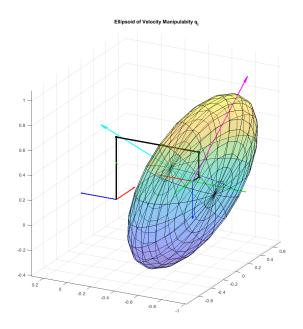
$$J = U\Sigma V^t \tag{5}$$

where the columns of U represent the principal axis of the ellipsoid and  $\Sigma$  contain the singular values  $\sigma_i$  of  ${}^0J_v(q)$ , which represent the length of the ellipsoid at each direction  $U_i$ .

It is also possible to calculate the preferred direction of the ellipsoid by taking the first column of U. In order to calculate the evelocity manipulability of a certain configuration, it is necessary to find the volume of the ellipsoid, which is done by the equation below.

$$W = \sqrt{\det(J(q)J^t(q))} = \prod_{i=1}^r \sigma_i \ge 0$$
(6)

#### 3.2.1 For configuration $q_i$



X
Y
Z
Velocity Ellipsold
axis ellipsold  $\sigma_1$ axis ellipsold  $\sigma_2$ axis ellipsold  $\sigma_3$ 

Figure 3.1 – Ellipsoid of Velocity Manipulability for  $q_i$ 

$$W = 0.1116$$
 and  $U_1 = [-0.7114, -0.0975, 0.6959]^T$ 

#### 3.2.2 For configuration $q_f$

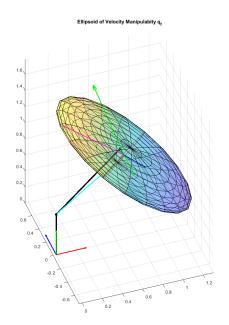




Figure 3.2 – Ellipsoid of Velocity Manipulability for  $q_f$ 

$$W = 0.0590$$
 and  $U_1 = [0.3760, -0.3263, 0.8715]^T$ 

# 4 Inverse geometric model

#### 4.1 Q8. Joint configuration by inverse geometric model

In this question the inverse geometrical model will be analysed by which it is possible to find a joint configuration q that allows the robot to position its tool in a desired position X.

To do this two different methods will be studied, the Newton-Raphson method and the Gradient method. The first one uses a Taylor series approximation to find the best solution. The second aims to minimize an objective-function H(q).

To implement both methods the function  $q*=ComputeIGM(X_d,q_0,k_{max},\epsilon_x)$  was developed based on the pseudo-algorithm illustrated by figure 4.1.  $X_d$ ,  $q_0$ ,  $k_max$  and  $\epsilon_x$  represent, respectively, the desired task position, the initial configuration, the maximum number of iterations and the norm of the tolerated Cartesian error. The last two parameters are used as stop criteria.

To study the performance of this model, two sets of parameters were proposed as seen below :

1. 
$$X_d = X_{di} = (-0.1, -0.7, 0.3)^t, q_0 = [-1.57, 0.00, -1.47, -1.47, -1.47, -1.47, -1.47, -1.47], k_{max} = 100, \epsilon_x = 1 \text{mm}$$

2. 
$$X_d = X_{df} = (0.64, -0.10, 1.14)^t, q_0 = [0, 0.80, 0.00, 1.00, 2.00, 0.00], k_{max} = 100, \epsilon_x = 1 \text{mm}$$

$$k \leftarrow 0$$

$$\text{while } \|J^t\left(q_k\right)\| > \epsilon \text{ do}$$

$$k \leftarrow k+1$$

$$\Rightarrow \text{ Case of the Gradient-based method}$$

$$q_k \leftarrow q_{k-1} + \alpha J^t\left(q_{k-1}\right)\left[X_d - f(q_{k-1})\right]$$

$$\Rightarrow \text{ Case of the Newton-Raphson-based method}$$

$$q_k \leftarrow q_{k-1} + J^{-1}\left(q_{k-1}\right)\left[X_d - f(q_{k-1})\right]$$

$$\text{end while}$$

$$q^* \leftarrow q_{k+1}$$

$$\text{return } (q^*)$$

FIGURE 4.1 – Pseudo-algorithm for computing the inverse geometric model

The following table shows the results for each case using both methods mentioned earlier.

Configuration	${f Method}$	Error (mm)	CPU time (ms)
$\mathrm{q}_i$	Newton-Raphson	$5.59 \cdot 10^{-4}$	6.45
$\mathbf{q}_i$	Gradient	$9.23 \cdot 10^{-4}$	7.79
$\mathbf{q}_f$	Newton-Raphson	$4.16 \cdot 10^{-4}$	10.98
$q_f$	Gradient	$57.00 \cdot 10^{-4}$	19.50

Table 3 – Geometric parameters of the robot

From the results obtained, it can be seen that for both cofigurations the Newton-Raphson method had a lower error and a lower CPU time than the gradient method. Thus, for the development of the following questions, the method used to calculate the Inverse Geometric Model will be that of Newton-Raphson.

# 5 Inverse kinematic model

# 5.1 Q9. Following a trajectory with inverse kinematic model

The next step is to impose a desired trajectory to be followed by the end effector. In this case, we will impose as trajectory a straight line from the initial position,  $X_{di}$ , to the final position,  $X_{df}$ . The rectilinear motion is carried out at a constant speed V = 1m/s and is sampled at period  $T_e = 1 ms$ .

The initial condition and final condition are:

- 1.  $X_{di} = (-0.10, -0.70, 0.30)^t$ 2.  $X_{df} = (0.64, -0.10, 1.14)^t$
- The trajectory is discretized in n points with a corresponding desired position Xdk.

The discretization is done as follows:

$$temp = \frac{||X_{df} - X_{di}||}{V}$$

$$n = ceil(temp/Te)$$

Then, we determine the setpoints values  $q_{dk}$  utilizing the function  $q^* = \text{ComputeIGM}(X_d, q_0, k_{max}, \epsilon_x)$ , developed in question 8. We utilize as the initial approximation for  $q_{dk}$ ,  $q_0 = q_{d(k-1)}$ .

The values computed for  $q_{dk}$  are saved in a vector and the sequence of positions of the frame of the end frame was plotted. The figure below shows the trajectory performed:

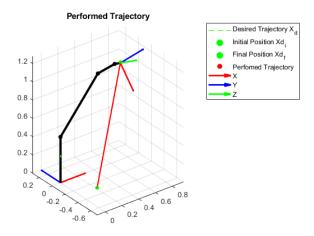


FIGURE 5.1 – Trajectory of the end effector

# 5.2 Q10. Temporal evolution of the joint variables

In this question, the temporal evolution of each joint variables  $q_1$  to  $q_6$  has been plotted, the maximum values  $(q_{max})$  and minimum values  $(q_{min})$  limitations for each one were overlayed in the plot.

$$q_{min} = [-\pi, -\frac{\pi}{2}, -\pi, -\pi, -\frac{\pi}{2}, -\pi]^t$$
 
$$q_{max} = [0, \frac{\pi}{2}, 0, \frac{\pi}{2}, \frac{\pi}{2}, \frac{\pi}{2}]^t$$

The following figure shows the temporal evolution of each variable.

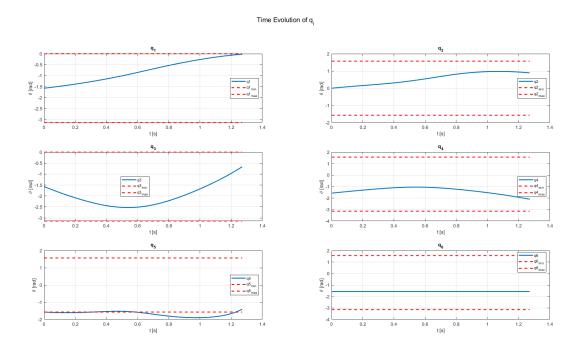


Figure 5.2 – Temporal evolution of  $q_i$ 

#### 5.3 Q11. Secondary task with inverse kinematic model

Given that in the previous question we saw that the angular limitations of each board were not very well respected, it is now desired to compute the inverse cinematic model taking into account a secondary task. Thus, a new function ComputeIKMlimits( $X_{di}$ ,  $X_{df}$ , V, Te,  $q_i$ ,  $q_{min}$ ,  $q_{max}$ ), based on the one created for question Q9, would limit the joints to approach the maximum and minimum values allowed. To that end, the following equation can be used to find the solution to a problem of the reverse kinematic model of a redundant case:

$$\dot{q}^* = J^{\#} \dot{X}_d + (I_n - J^{\#} J) \dot{q}_0 \tag{7}$$

where the first term of the sum is the particular solution whereas the second is the orthogonal projection of  $\dot{q}_0$  in the null space (represents the internal motion in the joint space).

To calculate the preferred velocity, the projected gradient technique was used and is expressed by the equation below :

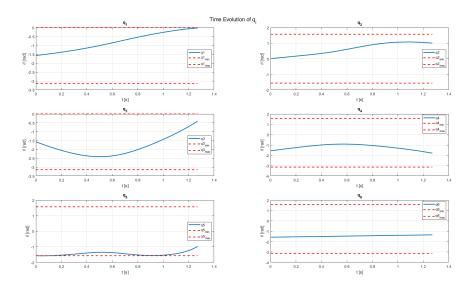
$$\dot{q}_0 = -\alpha \nabla_q H(q) = \begin{pmatrix} \frac{\partial H}{\partial q_1} \\ \vdots \\ \frac{\partial H}{\partial q_n} \end{pmatrix}$$
(8)

where H(q) is the objective function and  $\alpha$  is the tradeoff between the minimisation objectives of  $\frac{1}{2} \|\dot{q}\|^2$  and H(q).

Since, in this case, it is desired to limit the maximum and minimum angular rotation of the joints during the arm manipulation, the following objective function was used:

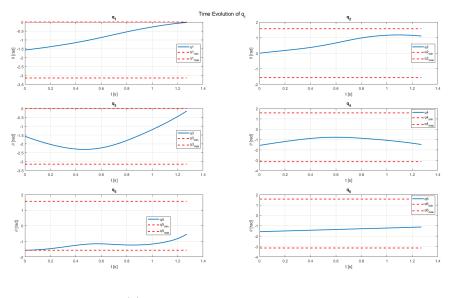
$$H_{lim.}(q) = \sum_{i=1}^{n} \left(\frac{q_i - \overline{q}_i}{q_{max} - q_{min}}\right)^2 \quad \text{where} \quad \overline{q}_i = \frac{q_{max} - q_{min}}{2} \tag{9}$$

By applying this new function to the same case performed in the previous question, new curves are obtained with respect to the angles of each joint in time. These curves can be seen in the figure below.



(a) Trade-off  $\alpha = 0.5$ , error  $\epsilon = 1.46$ 

FIGURE 5.3 – Time Evolution for the angle of the joints for two different configurations



(b) Trade-off  $\alpha=1.0,\,{\rm error}\ \epsilon=3.40$ 

FIGURE 5.3 – Time Evolution for the angle of the joints for two different configurations

The error between the achieved trajectory and the target trajectory was calculated for two different  $\alpha$  values and the trajectory of the robotic arm is shown in the figure below.

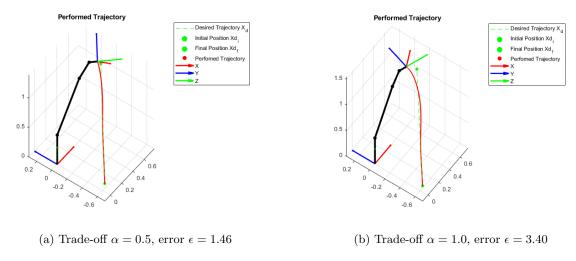


FIGURE 5.4 – Trajectory of the end effector with a imposed secondary task

By analysing figures 5 and 6, we can see the influence of the value of tradeoff on the results. While the tradeoff value is low ( $\alpha = 0.5$ ), the angular restrictions are not well respected but the new trajectory has a minor error compared to the target trajectory. On the other hand, for the tradeoff = 1, the error between the trajectories increases but the angular variables of each joint do not exceed the established limits.

# 6 Dynamic model

Now it is proposed to study the dynamic model of the robotic arm. The matrix form of the inverse dynamic model for rigid robot manipulator is presented below:

$$A(q)\ddot{q} + C(q,\dot{q})\dot{q} + G(q) + \Gamma_f(\dot{q}) = \Gamma \tag{10}$$

where A(q) is the inertia matrix,  $C(q, \dot{q})\dot{q}$  is the vector of joint torques due to the Coriolis and centrifugal forces, G(q) is the vector of joint torques due to gravity,  $\Gamma_f(\dot{q})$  is the vector of joint friction torques,  $\Gamma$  is the vector of the joint torques and q,  $\dot{q}$  and  $\ddot{q}$  are respectively the vectors of joint positions, velocities and accelerations

## 6.1 Q12. Velocity of the center of mass and rotation speed of all the rigid bodies

First of all, the velocity of the centre of mass and the speed of rotation of each rigid body of the system are calculated. To do this a new function ComputeJacGi( $\alpha$ ,d, $\theta$ ,r, $x_G$ , $y_G$ , $z_G$ ) has been created. Within this function the distance between the centre of mass of a rigid body and the frame  $R_0$  is first calculated and then the Jacobian matrix of this position is computed. Then from the Jacobian matrix the velocity of the centre of mass and the speed of rotation of the target body are obtained following the same method as in question  $\mathbf{Q6}$ .

$${}^{0}J_{Gi}(q) = \begin{bmatrix} {}^{0}J_{v_{Gi}}(q) \\ {}^{0}J_{\omega_{Gi}}(q) \end{bmatrix}$$
 (11)

## 6.2 Q13. Inertia Matrix of the robot

A function to calculate the inertial matrix of the robot was then created. In order to perform its calculation, the following formula was applied.

$$A(q) = \sum_{i=1}^{N} (m_i \, {}^{0}J_{Gi}^{t}(q) \, {}^{0}J_{Gi}(q) + \, {}^{0}J_{\omega i}^{t}(q) \, {}^{0}I_i \, {}^{0}J_{\omega i}(q))$$

$$(12)$$

where  ${}^{0}J_{Gi}$  is the velocity of the centre of mass of the body,  ${}^{0}J_{\omega i}$  is the speed of rotation of the body,  $m_{i}$  is the mass of the body and  ${}^{0}I_{i}$  is the inertia tensor evaluated at the centre of mass of the body transported to the frame  $R_{0}$ .

To calculate the matrix  ${}^{0}I_{i}$ , the Generalized Huygens theorem was used to first calculate  ${}^{0}I_{Gi}$ . Equation 13 represents this theorem.

$$I_{Oi} = I_{Gi} + m_i \begin{bmatrix} XX_i & XY_i & XZ_i \\ YX_i & YY_i & YZ_i \\ ZX_i & ZY_i & ZZ_i \end{bmatrix}$$

$$(13)$$

Then the  $I_{Gi}$  matrix was transported to the frame  $R_0$ ,  ${}^0I_i = R_0I_{Gi}R_0^T$ . Finally the motor inertia of each actuator calculated by  $r_{red}^2 * J_m$  is added the diagonal of matrix A(q).

#### 6.3 Q14. Lower and upper bounds of the inertia matrix

Since it is possible to calculate the inertial matrix of the robot for a given configuration of joint angles q, an A-bounds ( $q_{min}$ ,  $q_{max}$ , n-joints) function has been developed to verify the maximum and minimum value of A within the angular restrictions  $q_{min}$  and  $q_{max}$ . Thus A(q) can be defined as follows:

$$\mu_1 \le A(q) \le \mu_2$$

As the robot is composed only by revolute joints,  $\mu_1$  and  $\mu_2$  values are constant and do not depend on the joint configuration. To find these boundary values, a discretization is made for each joint angle  $q_i$  of the

robot, within the angle restrictions, and the value of A(q) is calculated for several combinations of joint angle configurations.

The minimum and maximum values of A(q) are respectively:

$$\mu_1 = 0.0589$$
 and  $\mu_2 = 10.1985$ 

# 6.4 Q15. Gravitational torque

The next step is to calculate the torque due to the gravitational force G(a) by function ComputeGravTorque(q). Through the gradient of the potential energy  $E_p$  due to gravitational force, the gravitational torque can be expressed by the following equation.

$$G(q) = -\binom{0}{v_{G_1}} \binom{t}{v_{G_1}} \binom{t}{v_{G_1}} \binom{t}{v_{G_6}} m_6 g \tag{14}$$

where  ${}^{0}J_{v_{Gi}}$  is the velocity of the centre of mass of the body,  $m_{i}$  is the mass of the body and g is the gravity vector.

## 6.5 Q16. Upper bound of the gravitational torque

In this question the maximum value of gravitational torque was checked among the various possibilities of angular configurations q of the robot joints. This was done through the function G\_bound( $q_{min}$ ,  $q_{max}$ ,  $n\_joints$ ) and the method used was the same as the function A\_bounds( $q_{min}$ ,  $q_{max}$ ,  $n\_joints$ ). As a result, the maximum value of G(q) = 116.8160 was obtained.

## 6.6 Q17. Inverse dynamic model

Then the last function  $ComputeFrictionTorque(\dot{q})$  was created. It returns the torque due to friction of the joints, which can be calculated from equation 15.

$$\Gamma_{f_i}(q_i) = diag(\dot{q}_i)F_{vi} \tag{15}$$

where  $\Gamma_{f_i}(q_i)$  is the torque due to friction,  $\dot{q}_i$  is the velocity vector of the joint i and  $F_{vi}$  is the joint viscous frictions.

Now, from all the functions created above, together with function  $ComputeCCTorques(q, \dot{q})$  which calculates the torque of the joints due to the Coriolis and centrifugal effects, already provided, it was possible to develop in  $Simulink^{TM}$  the inverse dynamic model (equation 10) which has as input the total torque of the system applied by the motors and as output the vector acceleration of the joints. The block built can be seen in the figure below.

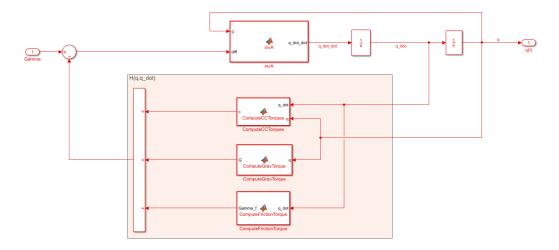


Figure 6.1 – Dynamic System Block Diagram

# 7 Trajectory generation in the joint space

# 7.1 Q18. Minimal time to follow trajectory

Given a starting configuration  $q_{d_i}$  and a final configuration  $q_{d_f}$  it is desired to calculate the minimum time  $t_f$  required to follow a polynomial trajectory of degree 5 in the joint space, performed at zero initial and final velocities and accelerations. The time is sampled at a period  $T_e = 1$ ms. Both initial and final configurations are as follow:

$$\begin{aligned} & - & q_{d_i} = [-1.00, \, 0.00, \, -1.00, \, -1.00, \, -1.00, \, -1.00]^t \text{ rad} \\ & - & q_{d_f} = [\, 0.00, \, 1.00, \, 0.00, \, 0.00, \, 0.00, \, 0.00]^t \text{ rad} \end{aligned}$$

To calculate the angular configuration of the robot as a function of the time q(t), the following expressions can be used :

$$q(t) = q_{d_i} + r(t)D \tag{16}$$

$$r(t) = 10(\frac{t}{t_f})^3 - 15(\frac{t}{t_f})^4 + 6(\frac{t}{t_f})^5$$
(17)

where  $D = q_{d_f} - q_{d_i}$ 

Taking into account only the  $k_{a_j}$  vector of maximum acceleration of the joints (calculated using equation 19), the minimum time  $T_{f_i}$  of the  $j^{th}$  joint to reach the final configuration is given by the equation below.

$$t_{f_i} = \sqrt{\frac{10 \ |D|}{\sqrt{3k_{a_j}}}} \tag{18}$$

$$k_{a_j} = \frac{\tau_{max,j} \ r_{red,j}}{\mu_2} \tag{19}$$

where  $\tau_{max,j}$  is the maximal torque of joint j and  $r_{red,j}$  is the reduction factor of joint j.

Applying all the values provided by the problem in the above equations results in the minimum time  $t_{f_i} = 0.4102 \ s$ .

# 7.2 Q19. Generating a desired joint trajectory point

Now through  $Simulink^{TM}$  the previous equations are implemented within an function that generates the trajectory  $q_c = \text{GenTraj}(q_{d_i}, q_{d_f}, t)$ , having as input the initial and final configuration of the joints and the minimum time. The diagram created can be seen as the following.

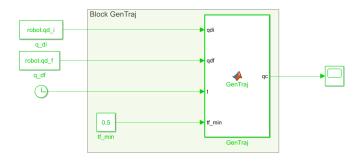


Figure 7.1 – Trajectory Generation Block Diagram

Using  $q_{d_i}$  and  $q_{d_f}$  from the previous question and  $t_f = 0.5$ , the angles of each joint j as a function of time, for the trajectory found, are presented below.

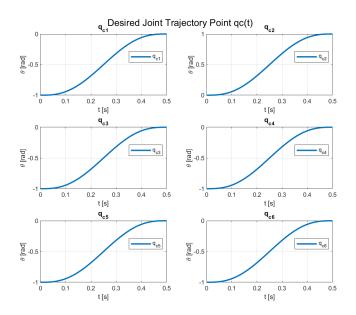


FIGURE 7.2 – Evolution of Joint Trajectories  $q_i(t)$ 

# 7.3 Q20. P.D. controller with gravity compensation

Finally, in order to finish the robotic arm project, a block to perform the position control is added to the system. This is done through a P.D controller with gravitational compensation, and can be described by the following expression :

$$\Gamma = K_p(q_d - q) + K_d(\dot{q}_d - \dot{q}) + \hat{G}(q)$$
(20)

The controller block built in  $Simulink^{TM}$  is schematized below.

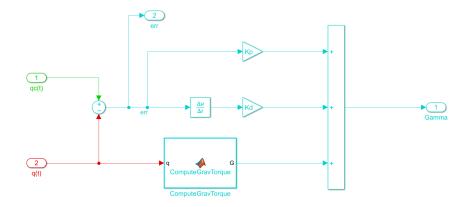


Figure 7.3 – Control Block Diagram

Now, by linking this block to the inverse dynamic model block and the trajectory generation block, we obtain the following system :

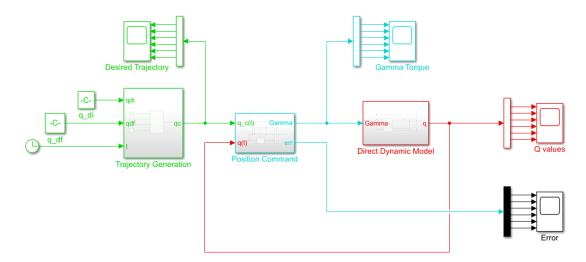


Figure 7.4 – Position-Control block Diagram

Through the analysis of velocity, acceleration, error and admissible torques, it was possible to define the values of  $K_d$  and  $K_p$  of the controller as the following :

- $-- \ q_{d_i} = [5000,\, 8000,\, 5500,\, 800,\, 5500,\, 2500]^t$
- $-- \ q_{d_f} = [1000, \, 1000, \, 500, \, 100, \, 100, \, 500]^t$

With the implementation of these values on the controller it was possible to obtain the graphs below regarding the temporal evolution of joint trajectories  $q_i(t)$ , tracking errors e (t), and control joint torques  $\Gamma_i(t)$  of each of the robot joints.

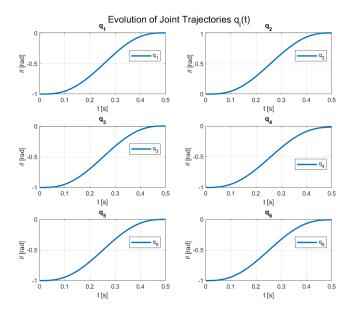


Figure 7.5 – Evolution of Joint Trajectories  $q_i(t)$ 

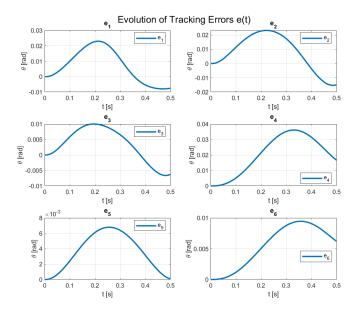


Figure 7.6 – Evolution of tracking errors  $e_i(t)$ 

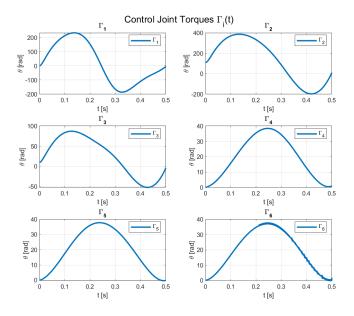


FIGURE 7.7 – Evolution of Control Joint Torques  $\Gamma_i(t)$ 

From the figure 7.6, it is possible to visualize that the joints are able to follow the desired joint trajectory  $q_c$  with a tracking error below the limits of the 0.05, for the Kp and Kd chosen.

Regarding the maximum torque, we see that the joints are all below the limits of the maximum torque from the motor of each joint. To verify that, it is possible to calculate the torque of the motor after the reduction and compare with the values obtained to the torque command by the control parameters. The torque limits of the motor after the reduction is given by:  $\Gamma_{max} = [500, 500, 500, 350, 350, 350]^T$  and from the figure 7.7, we verify that for all the joints the requested torque is below the limits.

# A Appendix - MatLab functions

#### A.1 Main Function

```
clear all
    close all
3
    clc
    addpath('functions')
    5
   %% Problem specifications
9
   10
11
   12
13
   14
15
        = 0.7:
16
        = 0.5:
17
   r1
        = 0.2;
18
   r4
   r_e = 0.1;
19
20
   21
22
    % Analyse Parameters
   23
24
25
   % Joint configuration qi
   qi = [-pi/2, 0, -pi/2, -pi/2, -pi/2, -pi/2];
26
27
   \% Joint configuration qf
28
29
   qf = [0, pi/4, 0, pi/2, pi/2, 0];
30
   % Joint velocity q_{dot} = [0.5, 1, -0.5, 0.5, 1, -0.5];
31
32
33
   34
   %% Question 2 - Robot Reference Parameters
    36
   robot.qmin = [-pi, -pi/2, -pi, -pi, -pi/2, -pi]';
robot.qmax = [ 0, pi/2, 0, pi/2, pi/2, pi/2]';
38
    robot.n_joints = 6;
    robot.n_frames = 7;
    robot.alpha = [0, pi/2, 0, pi/2, -pi/2, pi/2, 0]';
robot.d = [0, 0, d3, 0, 0, 0, 0]';
robot.theta = [0, 0, 0 + pi/2, 0, 0, 0, 0]';
                = [r1, 0, 0, r4, 0, 0, r_e]';
    \% Mass of each arm of the robot
48
    robot.m = [15.0, 10.0, 1.0, 7.0, 1.0, 0.5];
50
    % Robot's Joints Inertia Matrix
51
    robot.I = zeros(3,3,robot.n_joints);
52
53
    robot.I(:,:,1) = [0.80, 0.00, 0.05;
                       0.00, 0.80, 0.00;
                                             % Inertial tensor of the body 1
54
                       0.05, 0.00, 0.10]; % in the R1 frame[kg*m^2]
55
56
    robot.I(:,:,2) = [0.10, 0.00, 0.10;
57
                       0.00, 1.50, 0.00; % Inertial tensor of the body 2 0.10, 0.00, 1.50]; % in the R2 frame[kg*m^2]
58
59
60
   robot.I(:,:,3) = [0.05, 0.00, 0.00;
61
                       0.00, 0.01, 0.00;
0.00, 0.00, 0.05];
                                              % Inertial tensor of the body 3
62
                                              % in the R3 frame[kg*m^2]
63
64
    robot.I(:,:,4) = [0.50, 0.00, 0.00;
65
                       0.00, 0.50, 0.00; % Inertial tensor of the body 4 0.00, 0.00, 0.05]; % in the R4 frame[kg*m^2]
66
67
68
    robot.I(:,:,5) = [0.01, 0.00, 0.00;
69
                       0.00, 0.01, 0.00; % Inertial tensor of the body 5 0.00, 0.00, 0.01]; % in the R5 frame[kg*m^2]
70
71
72
73
    robot.I(:,:,6) = [0.01, 0.00, 0.00;
                       0.00, 0.01, 0.00; % Inertial tensor of the body 6 0.00, 0.00, 0.01]; % in the R5 frame[kg*m^2]
74
75
76
77
   % Vectors of Gi given in frame Ri
```

```
79
80
81
82
83
   % Moment of inertia of the actuator rotor [kg*m^2] robot.Jm = (10^-5)*[1; 1; 1; 1; 1; 1];
84
85
86
   % Reduction ratio
   robot.red = [100; 100; 100; 70; 70; 70]; % []
87
88
89
   % Joint viscous frictions
   robot.Fv = [10; 10; 10; 10; 10; 10]; % [N*m*(rad^-1)*s]
90
91
92
   % Maximal motor torques
   robot.tau_max = [5; 5; 5; 5; 5; 5]; % [N*m]
93
94
   95
96
97
98
99
   100
   %% Problem design
   102
   104
   %% Questions 3 and 4
   106
107
108
   % Analyse position
   theta_i = robot.theta + [qi;0];
109
   theta_f = robot.theta + [qf;0];
110
111
112
   % Transformation Matrix
   g_OEi = ComputeDGM(robot.alpha, robot.d, theta_i, robot.r);
113
114
   g_OEf = ComputeDGM(robot.alpha, robot.d, theta_f, robot.r);
115
   % Position Vector of End frame at qi and qf
116
   Pi = g_0Ei(1:3,4);
117
   Pf = g_0Ef(1:3,4);
118
119
   % Rotation Matrix of End frame at qi and qf
120
   Rot_i = g_0Ei(1:3,1:3);
121
   Rot_f = g_0Ef(1:3,1:3);
122
123
   % Unitary vector and angle of the transformation
124
   [phi_i,n_i] = inv_Rot(Rot_i);
125
   [phi_f,n_f] = inv_Rot(Rot_f);
126
127
128
   129
130
   %% Question 6
   131
132
   % Jacobian of End frame
133
134
   % Joint configuration qi
135
   J0_Ei = ComputeJac(robot.alpha, robot.d, theta_i, robot.r, Pi, robot.n_joints);
136
137
   % Joint configuration qf
138
   J0_Ef = ComputeJac(robot.alpha, robot.d, theta_f, robot.r, Pf, robot.n_joints);
139
140
   VO_OEi = JO_Ei*q_dot;
VO_OEf = JO_Ef*q_dot;
141
142
143
144
   145
146
   %% Question 7
147
   148
   \% Single value decomposition
149
   [U_i, Sig_i, V_i] = svd(J0_Ei(1:3,:));
[U_f, Sig_f, V_f] = svd(J0_Ef(1:3,:));
150
151
152
153
   % Velocity Manipulability
   Vel_Man_i = prod(diag(Sig_i));
Vel_Man_f = prod(diag(Sig_f));
155
157
   158
   %% Question 8
   161
162 % Joint configuration qi
```

```
165
166
       \mbox{\ensuremath{\mbox{\%}}} Joint configuration qf
       Xd_f = [0.64, -0.1, 1.14]';
q0_f = [0, 0.80, 0.00, 1.00, 2.00, 0.00]';
167
168
169
170
       % Max number of iterations
171
       k_max = 100;
172
       % Max error
       eps_max = 0.001;
173
174
       % IGM by Newton-Raphson Method
175
       [f_in, err_in, q_star_in] = ComputeIGM(robot, Xd_i, q0_i, k_max, eps_max, 'NewtonRaphson');
[f_fn, err_fn, q_star_fn] = ComputeIGM(robot, Xd_f, q0_f, k_max, eps_max, 'NewtonRaphson');
176
177
178
       % Gradient Method
179
        [f_ig, err_ig, q_star_ig] = ComputeIGM(robot, Xd_i, q0_i, k_max, eps_max, 'Gradient');
180
       [f_fg, err_fg, q_star_fg] = ComputeIGM(robot, Xd_f, q0_f, k_max, eps_max, 'Gradient');
181
182
       184
185
       %% Question 9 and 10
       186
       % Linear velocity
188
       V_d = 1;
       % Time sample
190
       Te = 0.001;
191
192
193
       % IKM
       [Xdk, thetadk, qdk] = ComputeIKM(robot, Xd_i, Xd_f, V_d, Te, qi, k_max, eps_max, 'NewtonRaphson');
194
195
196
       197
198
       %% Question 11
       199
200
       % IKMLimits for tradeoff = 0.5
201
       % tradeoff = 0.5;
202
       % [X_star, theta_star, q_star] = ComputeIKMLimits(robot, Xd_i, Xd_f, V_d, Te, qi, robot.qmin, robot.qmax,
203
                k_max, eps_max, tradeoff, 'NewtonRaphson');
       % err_tradeoff_05 = norm(X_star - Xdk);
204
205
       % IKMLimits for tradeoff = 1
206
        tradeoff = 1;
207
        [\texttt{X\_star}, \texttt{theta\_star}, \texttt{q\_star}] = \texttt{ComputeIKMLimits(robot, Xd\_i, Xd\_f, V\_d, Te, qi, robot.qmin, robot.qmax, quality of the property o
208
       k_max, eps_max, tradeoff, 'NewtonRaphson');
err_tradeoff_10 = norm(X_star - Xdk);
209
210
211
212
       213
       %% Question 12
       214
215
216
       Joint_test_JacGi= 6;
217
       \% Jacobian matrix of centre of mass of rigid body [Jv, Jw, p_cg_00, g] = ComputeJacGi(robot.alpha, robot.d, [qi;0] + robot.theta, robot.r, robot.cg,
218
219
               Joint_test_JacGi);
220
221
       222
223
       %% Question 13
       224
225
226
       % Inertia matrix
227
       Ai = ComputeMatInert(robot, qi);
228
        Af = ComputeMatInert(robot, qf);
229
230
231
       %% Question 14
232
       233
234
       % Min and Max values for the inertia matrix
236
        [mu_2,mu_1] = A_bounds(robot);
237
238
       239
240
       %% Question 15
       242
243 % Torque due to the gravity
```

```
244 | Gi = ComputeGravTorque(robot, qi);
       Gf = ComputeGravTorque(robot, qf);
245
246
247
248
       249
       %% Question 16
       250
251
       \mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath}\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ens
252
253
       gb = G_bounds(robot);
254
255
256
       257
       %% Question 17
       258
259
260
       % Torque due to joint friction
261
       Gamma_f = ComputeFrictionTorque(robot,q_dot);
262
263
       264
265
       %% Question 18
266
       267
       \mbox{\ensuremath{\mbox{\%}}} Initial and final cofigurations
       % [rad] Initial position
269
270
272
       % Minimal time to follow trajectory
273
       Dj = robot.qd_f - robot.qd_i;
       k_aj = robot.tau_max.*robot.red/mu_2;
274
       tf_min = max( sqrt( 10*abs(Dj)./(sqrt(3)*k_aj) ));
275
276
277
       278
279
       %% Question 19
       280
281
       % Trajactory generation
282
       qc = GenTraj(qdi, qdf,t, tf_min);
283
284
285
       286
287
       %% Question 20
       288
289
       % Control parameters
290
       Kp = [5000; 8000; 5500; 800; 5500; 2500];
Kd = [1000; 1000; 500; 100; 100; 500];
291
292
293
294
       295
296
       %% Simulink
       297
298
       open_system('main_simulink');
299
300
       sim('main_simulink.slx');
301
302
303
304
       305
       306
       %% Plots
       307
308
       309
310
       %%
311
       figure(1)
312
       PlotFrame(g_OEi) % Q5
313
       hold on
       \tt create\_ellipsoide(Pi,Sig\_i,U\_i)~\%~Q7
314
       PlotRobot(robot,qi)
       legend('X','Y','Z','Velocity Ellipsoid','axis ellipsoid \sigma_1','axis ellipsoid \sigma_2','axis
ellipsoid \sigma_3')
316
       title('Ellipsoid of Velocity Manipulabity q_i')
317
318
       %%
320
       figure(2)
321
       PlotFrame(g_OEf) % Q5
322
       create_ellipsoide(Pf,Sig_f,U_f) % Q7
325 PlotRobot(robot,qf)
```

```
legend('X','Y','Z','Velocity Ellipsoid','axis ellipsoid \sigma_1','axis ellipsoid \sigma_2','axis
ellipsoid \sigma_3')
326
     title('Ellipsoid of Velocity Manipulabity q_f')
327
328
329
    %% Question 9
330
331
    figure(3)
332
     title('Performed Trajectory qdk')
    AnimationTrajectory(robot, thetadk, qdk, Xd_i, Xd_f, Xdk, 'q9.avi')
333
334
335
336
    %% Question 10
337
    figure (4)
338
     suptitle('Time Evolution of q_i')
    JointTemporalPlot(robot.qmin, robot.qmax, Te, qdk)
339
340
    %% Question 11
341
342
    figure (5)
343
    suptitle('Time Evolution of q_i with Secondary Task')
344
    JointTemporalPlot(robot.qmin, robot.qmax, Te, q_star)
    title('Performed Trajectory with Secondary Task')
     AnimationTrajectory(robot,theta_star,q_star,Xd_i,Xd_f,X_star,'q11.avi')
348
     figure(7)
350
    suptitle('Desired Joint Trajectory Point qc(t)')
    simdata_plot(sim_qc, 'q_c_')
352
    %saveas(gcf,'q19_qc.png')
353
354
    %% Q20
355
    figure(8)
356
    suptitle('Control Joint Torques \Gamma_i(t)')
simdata_plot(sim_Gamma, '\Gamma_')
357
358
    %saveas(gcf,'q20_Gamma.png')
359
360
    figure(9)
361
    suptitle('Evolution of Joint Trajectories q_i(t)')
362
    simdata_plot(sim_q, 'q_')
363
    %saveas(gcf,'q20_q.png')
364
365
    figure(10)
366
    suptitle('Evolution of Tracking Errors e(t)')
367
    simdata_plot(sim_error, 'e_')
368
    %saveas(gcf,'q20_error.png')
369
370
    figure(11)
371
    suptitle('Evolution of q_i(t) and q_c(t)')
simdata_plot(sim_qc, 'qc_')
372
373
374
    hold on
    simdata_plot(sim_q, 'q_')
375
376
    %saveas(gcf,'q20_qc_q.png')
```

Code 1 - Main

# A.2 Direct geometric model

#### A.2.1 TransformMatElem

```
function g = TransformMatElem(alpha, d, theta, r)
2
   %TransformMatElem
   % Calculates the transformation matrix of a given grame
3
   % with respect to the reference frame
5
6
    g = [cos(theta),
                                     -sin(theta),
                                                                          Ο,
         cos(alpha)*sin(theta), cos(alpha)*cos(theta),
sin(alpha)*sin(theta), sin(alpha)*cos(theta),
                                                                     -sin(alpha), -r*sin(alpha);
                                                                       cos(alpha), r*cos(alpha);
                      Ο,
9
                                                   Ο,
                                                                          Ο,
                                                                                         1];
10
    end
```

Code 2 - TransformMatElem

#### A.2.2 ComputeDGM

```
function [g_ON] = ComputeDGM(alpha, d, theta, r)
   2
                       %ComputeDGM
  3
                     % Calculates the succesive transformations from the
   4
                     \mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath{\mbox{\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremath}\ensuremat
   5
   6
                     \% Returns the transformation matrix g_ON, 4x4.
                                                 n_frames = length(alpha);
   9
                                                 g_0N = eye(4);
10
                                                  for i=n_frames:-1:1
                                                                            g = TransformMatElem(alpha(i), d(i), theta(i), r(i));
11
12
                                                                            g_0N = g*g_0N;
                       end
```

Code 3 - ComputeDGM

## A.3 Direct kinematic model

#### A.3.1 ComputeJac

```
function J = ComputeJac(alpha, d, theta, r, p_e, n_joints)
   %ComputeJac
2
   % Calculates the jacobian matrix of a given point
3
   % Returns the jacobian matrix of size n_joints x n_joints.
5
6
        J = zeros(6,n_joints);
        g_0N = eye(4);
8
        % Khalil-Kleinfinger Convention
9
        Zi = [0; 0; 1];
10
11
        for i=1:n_joints
12
            g = TransformMatElem(alpha(i), d(i), theta(i), r(i));
13
            g_0N = g_0N*g;
14
15
16
            % Revolution Joint
            R_0i = g_0N(1:3,1:3);

p_in = g_0N(1:3,4);
17
18
19
            p = p_e - p_in;
20
            J(:,i) = [cross(R_0i*Zi,p);
21
22
                       R_0i*Zi;];
23
        end
24
25
    end
```

Code 4 – ComputeJac

# A.4 Inverse geometric model

#### A.4.1 ComputeIGM

```
function [Xd_estimated, err, q_star] = ComtputeIGM(robot, Xd, q_0, k_max, eps_x, method)
    %ComputeIGM
   \% Calculates the estimated position from a desired position
   % and an initial joint configuration by an iterative method
   % Returns the estimated position, the error and the estimated
   % joint configuration
        % tic
        step_IGM = 2;
10
        it = 1;
        err = 5000;
12
        f = [0,0,0];
13
        while (it < k_max && err > eps_x)
14
            theta_0 = [q_0;0] + robot.theta;
15
             g = ComputeDGM(robot.alpha, robot.d, theta_0, robot.r);
16
             Xd_{estimated} = g(1:3,4);
17
             J_full = ComputeJac(robot.alpha, robot.d, robot.theta + q_0, robot.r, Xd);
J_full = ComputeJac(robot.alpha, robot.d, theta_0, robot.r, Xd, robot.n_joints);
18
19
             J = J_full(1:3,:);
20
```

```
21
            if strcmp(method, 'NewtonRaphson')
22
                J_star = pinv(J);
23
24
            end
25
            if strcmp(method, 'Gradient')
26
                J_star = step_IGM*J';
27
            end
            q_k = q_0 + J_star*(Xd-Xd_estimated);
28
            err = norm(Xd-Xd_estimated);
29
            it = it + 1;
30
31
32
            q_0 = q_k;
33
        end
34
35
36
        q_star = q_0;
37
        fprintf("Method %s: ",method); toc
38
    end
```

Code 5 – ComputeIGM

## A.4.2 ComputeIKM

```
function [Xdk, thetadk, qdk] = ComputeIKM(robot, Xd_i, Xd_f, V, Te, q_i, k_max, eps_x, method)
   %ComputeIKM
    % Calculates the best points and joint configurations of the desired
   % trajectory
        delta_X = Xd_f - Xd_i;
        dist = norm(delta_X);
        u_x = (Xd_f - Xd_i)/dist;
9
10
        temp = dist/V;
        it= ceil(temp/Te);
11
12
        Xdk = zeros(length(Xd_i),it);
        qdk = zeros(length(q_i),it);
13
14
        % iter 1
15
        Xdk(:,1) = Xd_i;
16
        qdk(:,1) = q_i;
17
18
        for i=2:it-1
19
            Xdk(:,i) = Xdk(:,i-1) + (V*Te)*u_x;
20
            [~, ~, qdk(:,i)] = ComputeIGM(robot, Xdk(:,i), qdk(:,i-1), k_max, eps_x, method);
21
22
23
24
25
        Xdk(:,it) = Xd_f;
        [~, ~, qdk(:,it)] = ComputeIGM(robot, Xdk(:,it), qdk(:,it-1), k_max, eps_x, method);
26
27
        thetadk = [qdk;zeros(1,it)] + robot.theta;
28
29
    end
30
```

Code 6 - ComputeIKM

#### A.4.3 ComputeIKMlimits

```
k_max, eps_x, tradeoff, method)
   %ComputeIKMLimits
   \mbox{\ensuremath{\mbox{\%}}} Calculates the best points and joint configurations of the desired
3
   \mbox{\ensuremath{\mbox{\%}}} trajectory while respecting the angle restrictions qmax and qmin
4
       n = length(q_i);
delta_X = Xd_f - Xd_i;
5
6
       dist = norm(delta_X);
u_x = (Xd_f - Xd_i)/dist;
8
9
10
       temp = dist/V;
11
       it= ceil(temp/Te);
12
       t_lim = 0:Te:temp;
13
       Xdk_dot = V*u_x;
14
       q_star= zeros(length(q_i),it);
15
       q_star_dot = zeros(length(q_i),it);
16
       X_star = zeros(3,it);
17
       % iter 1
18
```

```
[Xdk, the tadk, qdk] = ComputeIKM(robot, Xd_i, Xd_f, V, Te, q_i, k_max, eps_x, method);
19
20
        qi_dash = (q_max + q_min)/2;
21
22
        for k=1:it
23
            dH_{\lim} = 2*(qdk(:,k)-qi_dash)./((q_max - q_min).^2);
            q0_dot = -tradeoff*dH_lim;
24
25
26
            J = ComputeJac(robot.alpha, robot.d, thetadk(:,k), robot.r, Xdk(:,k), robot.n_joints);
27
            rank_J(k) = rank(J);
28
            Jv = J(1:3,:);
29
            J_sharp = pinv(Jv);
30
            N_{j} = eye(length(Jv(1,:))) - (J_sharp*Jv);
31
            q_star_dot(:,k) = J_sharp*Xdk_dot + N_j*q0_dot;
32
33
34
        for i=1:n
35
            q_star(i,:) = cumtrapz(t_lim,q_star_dot(i,:));
36
        end
37
38
        q_star(:,:) = q_star(:,:) + q_i;
        theta_star = [q_star; zeros(1, it)] + robot.theta;
40
41
42
            g = ComputeDGM(robot.alpha, robot.d, theta_star(:,k), robot.r);
            X_{star}(:,k) = g(1:3,4);
44
46
    end
47
```

Code 7 - ComputeIKMlimits

# A.5 Inverse dynamic model

#### A.5.1 ComputeJacGi

```
function [Jv, Jw, p_cg_00, g] = ComputeJacGi(alpha, d, theta, r, cg, ji)
   %ComputeJacGi
2
   % Calculates the jacobian matrix of the centre of mass
3
   % of a rigid body
4
5
   % Returns the jacobian matrix of size ji x ji.
6
       alpha_cg = alpha(1:ji);
       d_cg = d(1:ji);
8
       theta_cg = theta(1:ji);
9
       r_cg = r(1:ji);
g = ComputeDGM(alpha_cg,d_cg,theta_cg,r_cg);
10
11
12
       p_cg_00 = g(1:3,:)*[cg(:,ji);1];
13
       J = zeros(6);
14
15
       J(:,1:ji) = ComputeJac(alpha_cg,d_cg,theta_cg,r_cg, p_cg_00,ji);
16
       Jv = J(1:3,:);
17
18
       Jw = J(4:6,:);
19
20
   end
```

Code 8 - ComputeJacGi

#### A.5.2 ComputeMatInert

```
function A = ComputeMatInert(robot, q)
   %ComputeMatInert
2
   % Calculates the inertia matrix of a given configuration
3
   % Returns the inertia matrix of size n_joints x n_joints.
6
       theta_0 = robot.theta + [q;0];
         theta_0 = robot.theta;
8
         theta_0(1:6) = theta_0(1:6) + q;
9
       A = zeros(6);
10
11
       for i=1:length(a)
12
13
           [Jvi, Jwi, p_cg_00, g] = ComputeJacGi(robot.alpha, robot.d, theta_0, robot.r, robot.cg, i);
14
```

```
IG\_i = robot.I(:,:,i) - robot.m(i)*skew(robot.cg(:,i));
15
             RO_i = g(1:3,1:3);
IO_i = RO_i*IG_i*RO_i';
16
17
             A = A + (robot.m(i)*(Jvi'*Jvi) + Jwi'*IO_i*Jwi);
18
19
         end
         A = A + diag(robot.red.^2.*robot.Jm);
20
21
22
23
    function M = skew(x)
24
            0, -x(3), x(2);
x(3), 0, -x(1);
25
         X = [
26
            -x(2), x(1), 0];
27
        M = -X * X;
28
```

Code 9 – ComputeMatInert

# A.5.3 ComputeGravTorque

```
function G = ComputeGravTorque(robot,q)
    %ComputeGravTorque
2
   % Computes the resistives torques in the joints due to % gravitational effects.
3
4
   \% Returns a vector 6x1, with the corresponding resisitve torque
5
6
        theta_0 = robot.theta + [q;0];
7
        G = zeros(length(q),1);
8
9
        for i=1:length(q)
10
             [Jv, ~] = ComputeJacGi(robot.alpha, robot.d, theta_0, robot.r, robot.cg, i);
G = G - robot.m(i)*(Jv'*robot.g);
11
12
13
14
15
    end
```

Code 10 - ComputeGravTorque

#### A.5.4 ComputeFrictionTorque

```
function Gamma_f = ComputeFrictionTorque(robot,q_dot)
%ComputeFrictionTorque
% Computes the Friction Torque from the viscous frictions
4 % and the current speed
5 % Returns a vectir Gamma_f 6x1.
6 Gamma_f = diag(q_dot)*robot.Fv;
8 end
```

Code 11 – ComputeFrictionTorque

#### A.5.5 GenTraj

Code 12 – GenTraj