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RESEARCH ARTICLE



# Decoupled Motion Planning of a Mobile Manipulator for Precision Agriculture

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## Abstract

Thanks to recent developments in service robotics technologies, precision agriculture is becoming an increasingly prominent research field, and several studies were made to present and outline how the use of mobile robotic systems can help and improve farm production. In this paper, the integration of a custom-designed mobile base with a commercial robotic arm is presented, showing the functionality and features of the overall system for crop monitoring and sampling. To this aim, the motion planning problem is addressed, developing a tailored algorithm based on the so-called manipulability index, that treats the base and robotic arm mobility as two independent degrees of motion; also developing an open source closed-form Inverse kinematics algorithm for the kinematically redundant manipulator. The presented methods and sub-system, even though strictly related to a specific mobile manipulator system, can be adapted not only to precision agriculture applications where a mobile manipulator is involved, but also to the wider field of assistive robotics.

## 1. Introduction

In the last years, Precision Agriculture (PA) has become a prominent research field, thanks to its significant impact for responsible water resource consumption [1], but also fertilizers, chemicals and pesticides reduction [2, 3]. The increasing population growth rate, that is expected to bring the global population to 9.7 billion by 2050 [4], is leading to a raise in food demand [5, 6, 7], especially in developing countries where land was withdrawn from cultivation for cities expansion [9], with a greater exploitation of the fields and mass crops. PA methods are nevertheless devoted to a data-driven approach for responsible and intelligent crops farming, thus pursuing the 2030 Agenda for Sustainable Development, especially the Sustainable Development Goal (SDG) 12 towards sustainable consumption and production patterns [8].

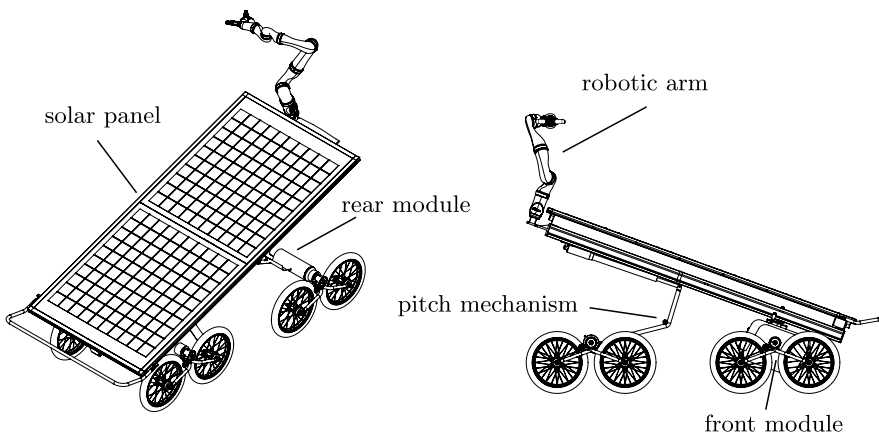
To this aim, a massive scientific activity about the implementation of novel technologies for crops farming was performed in the last decades, especially about integrated mobile systems that are capable of autonomous navigation while monitoring the crops [10, 11, 12], or even capable of helping farmers during their activities, e.g. trimming [13, 14, 15], spraying [16, 17] and harvesting [18, 19]. All the cited systems are briefly presented in Table 1 and their main features, e.g. the main purpose and manipulator type, are reported.

**Table 1.** *Mobile manipulators for precision agriculture.*

Study	Purpose	Target	Base	Locomotion	Arm
Strisciuglio N. et al. [13]	Gardening	Rose Bush	Commercial	Differential Drive	Commercial (6 d.o.f.)
Adamides G. et al. [16]	Spraying	Vineyard	Commercial	Differential Drive	Commercial (6 d.o.f.)
Feng Q. et al. [18]	Harvesting	Tomato fruit	Custom	Rail Guided	Commercial (4 d.o.f.)
Vinum project [14]	Pruning	Vineyard	Commercial	Legged	Commercial (7 d.o.f.)
De Preter et al. [19]	Harvesting	Strawberry (greenhouse)	Custom	Differential Drive	Custom
Botteril T. et al. [15]	Pruning	Vineyard	Custom	Winch driven	Commercial (6 d.o.f.)
Goričaneč J. et al. [17]	Spraying	Vineyard	Custom	Flipper-tracked	Commercial (7 d.o.f.)

To develop a novel UGV (Unmanned Ground Vehicle) for Precision Agriculture purposes, the researchers from Politecnico di Torino had developed the Agri.Q prototype, depicted in Fig. 1, which is a custom mobile manipulator mainly designed for grapevine cultivation [20, 21]. Agri.Q locomotion system is composed of two modules: the front and rear modules, both equipped with two drive motors and four wheels. While the significant number of wheels is fundamental to properly distribute the UGV weight on the ground, the all-wheels drive mode is fundamental to navigate in case of steep slopes or uneven terrains. The system is also provided with a photovoltaic panel, which allows the self-charging mode, and it can be properly oriented, thanks to a controlled pitch-roll motion, to exploit the solar radiation, thus maximizing the self-charging mode. Moreover, the integration with the Kinova Jaco2 commercial robotic arm makes the system capable of performing few tasks where the interaction with the environment is fundamental, e.g. crop sampling and spraying.

In Fig. 2 the Agri.Q High-Level architecture is presented. The electric energy is stored into a 7S14P Li-Po battery, that can be both re-charged by an external supply or the PV panels themselves. The rover can be directly commanded by a human being user through a remote control, whose receiver gives the

**Figure 1.** *The Agri.Q mobile manipulator for Precision Agriculture.*

input commands to the Low-Level micro-controller that drives the mobile base motors. Alternatively, the mobile base can operate in autonomous mode through the commands provided by an on-board computer, which can also control the robotic manipulator.

Apart from the novelty of the Agri.Q prototype, that was already discussed by the authors within the previously cited works, this article presents the motion planning pipeline that was developed to allow Agri.Q to perform simple crop sampling activities, exploiting both the mobile base and the arm mobility. Significant contributions of the paper are specified hereinafter:

- The implementation of a closed-form Inverse Kinematics algorithm that exploits the elbow redundancy for the Jaco2 manipulator is extensively presented and developed, and it is also provided by the authors as an open-source code for further research activities;
- The whole system motion planning pipeline is reported, presenting also how the decoupling of the base mobility and arm mobility can be used to place the target object of the pick-and-place task inside the area where the arm exhibits its highest manipulation capacities.

The present work is the extended version of the paper presented by the authors at the 31st International Conference on Robotics in the Alpe-Adria-Danube Region [22]. The following chapters are organized as follows:

- Section 2 presents the Inverse Kinematics of the manipulator and the estimation of the manipulation abilities with the use of a modified version of the manipulability index;
- Section 3 discusses the implementation of the arm kinematics within the whole system motion planning architecture.

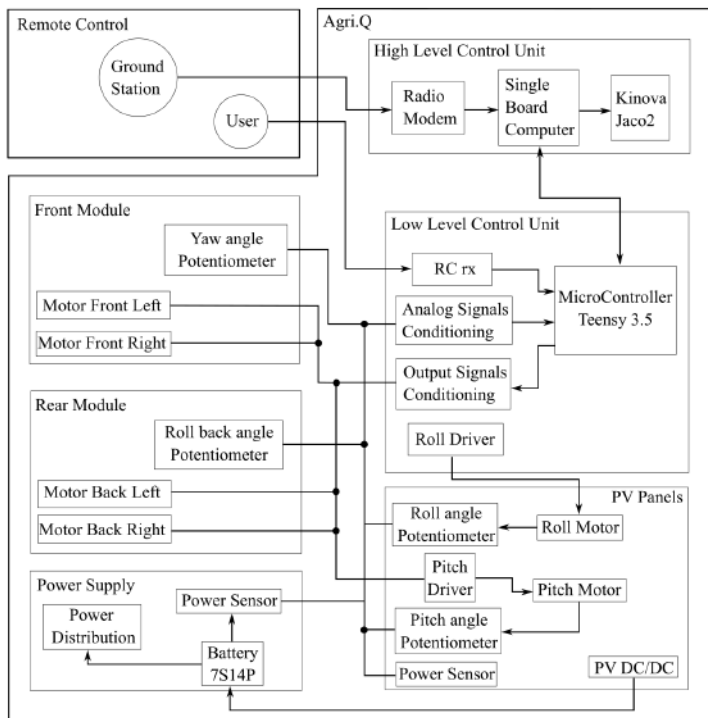


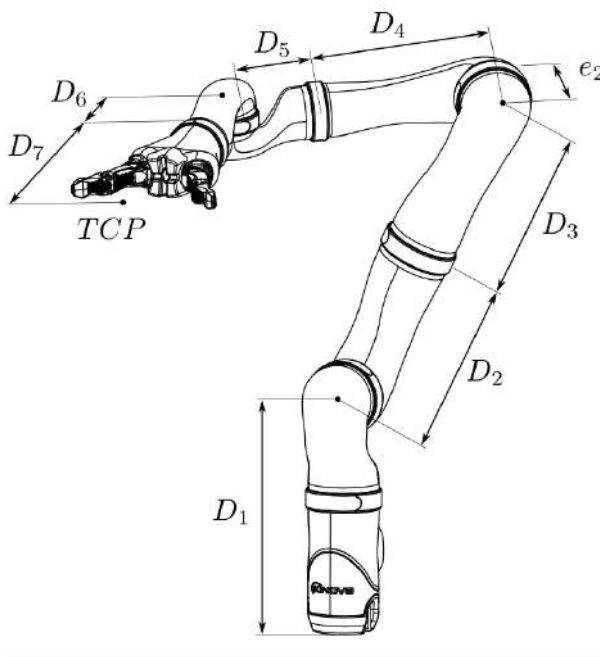
Figure 2. High-Level architecture of the Agri.Q mobile manipulator.

## 2. Jaco2 Manipulator Inverse Kinematics

The Kinova Jaco2 robotic manipulator can be modeled as a serial kinematic chain composed of eight links and seven revolute joints, where the first link is fixed to the Agri.Q rover and the last one represents the end-effector. In Fig. 3 the Jaco2 manipulator and its geometric properties are presented. It is worth underlining the elbow off-set  $e_2$  inducing a misalignment in the kinematic chain that will be discussed further. Since the arm should operate in a three dimensional workspace with an arbitrarily defined orientation, its seven degrees of freedom guarantee the kinematic redundancy of the system [24], thus, for each pose, there are  $\infty^1$  possible solutions to the inverse kinematics problem. The definition of the link frames was performed through the use of the Denavit-Hartenberg convention, even though the coefficient values were slightly modified to make them consistent when commanding the real manipulator, as can be verified by comparing the values in Tab. 2 with the parameters in the official documentation [23]. According to the official Kinova documentation, the fixed homogeneous transformation matrix

**Table 2.** Denavit-Hartenberg parameter values for the Jaco2 Manipulator for each  $i$ -th reference frame.

$i$	$a_i$	$\alpha_i$	$d_i$	$\theta_i$
1	0	$-\pi/2$	$-D_1$	$q_1$
2	0	$-\pi/2$	0	$q_2$
3	0	$-\pi/2$	$-(D_2 + D_3)$	$q_3$
4	0	$-\pi/2$	$-e_2$	$q_4$
5	0	$-\pi/2$	$-(D_4 + D_5)$	$q_5$
6	0	$-\pi/2$	0	$q_6$
7	0	$-\pi$	$-(D_6 + D_7)$	$q_7$



**Figure 3.** Kinova Jaco2 Manipulator and its fundamental geometric properties. The elbow off-set is labelled as  $e_2$ .

that sets  $\{0\}$  with respect to  $\{b\}$  is the following:

$$\mathbf{T}_0^b = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad (1)$$

The position and orientation of the frame fixed to the  $i$ -th link can be easily found with the composition rule of homogeneous transformation matrices:

$$\mathbf{T}_i^b(q_1, \dots, q_i) = \begin{bmatrix} \mathbf{R}_i^{i-1} & \mathbf{p}_i^{i-1} \\ \mathbf{0} & 1 \end{bmatrix} = \begin{bmatrix} r_i^{i-1}{}_{11} & r_i^{i-1}{}_{12} & r_i^{i-1}{}_{13} & p_i^{i-1}{}_x \\ r_i^{i-1}{}_{21} & r_i^{i-1}{}_{22} & r_i^{i-1}{}_{23} & p_i^{i-1}{}_y \\ r_i^{i-1}{}_{31} & r_i^{i-1}{}_{32} & r_i^{i-1}{}_{33} & p_i^{i-1}{}_z \\ 0 & 0 & 0 & 1 \end{bmatrix} = \mathbf{T}_0^b \mathbf{T}_1^0(q_1) \dots \mathbf{T}_i^{i-1}(q_i) \quad (2)$$

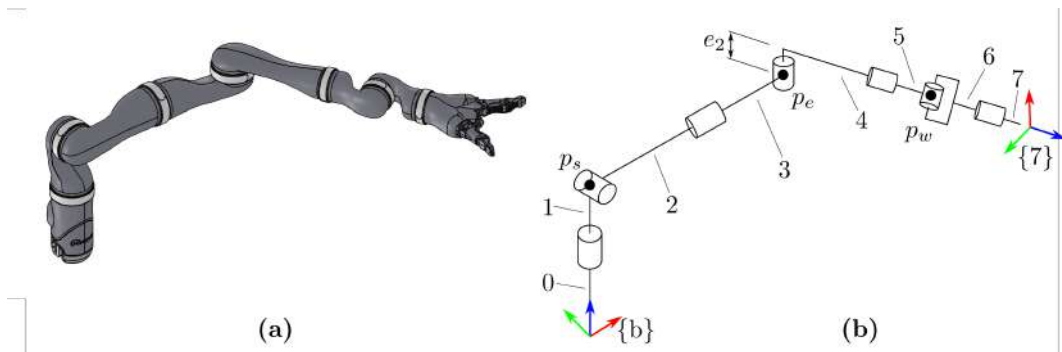
where  $q_i$  is the  $i$ -th generalized coordinate and  $T_i^{i-1}$  is the so-called homogeneous transformation matrix of frame  $i$  with respect to frame  $\{i-1\}$ , that can be computed as a function of the DH parameter and  $q_i$ :

$$\mathbf{T}_i^{i-1} = \begin{bmatrix} \cos(q_i) - \cos(\alpha_i) \sin(q_i) & \sin(\alpha_i) \sin(q_i) & a_i \cos(q_i) \\ \sin(q_i) & \cos(\alpha_i) \cos(q_i) & -\sin(\alpha_i) \cos(q_i) & a_i \sin(q_i) \\ 0 & \sin(\alpha_i) & \cos(\alpha_i) & d_i \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad (3)$$

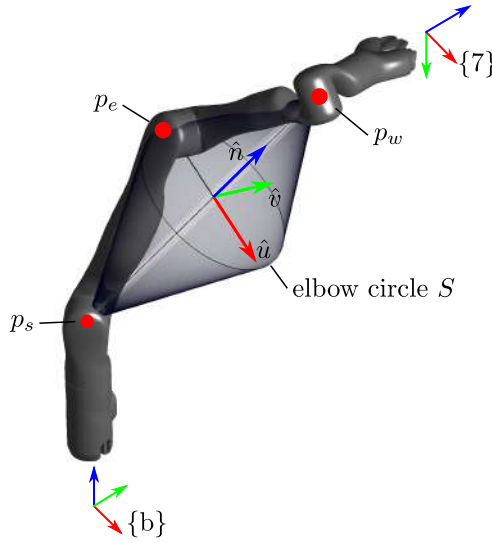
where, in the following, the compact notation  $s_i = \sin(q_i)$  and  $c_i = \cos(q_i)$  is also used.

### 2.1. Swivel Angle Approach

To face the Inverse Kinematics problem and compute the posture (or, in general, the postures) that is consistent with a specified end-effector pose, the serial kinematic chain that forms the manipulator can be represented as the composition of two spherical joints linked by a revolute joint. Indeed, as shown in



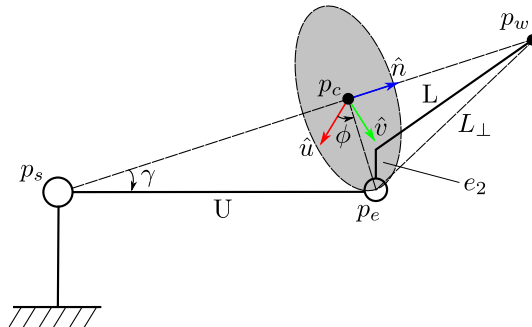
**Figure 4.** Modeling of the Kinova Jaco2 Manipulator as a composition of two spherical joints with an intermediate revolute joint. (a) 3D manipulator model, (b) Kinematic model.



**Figure 5.** Rotation of the elbow point  $p_e$  along the swivel circle  $S$  of radius  $R$ .

Fig. 4, the axes of the first three joints intersect at a single point named  $p_s$ , where  $s$  stands for "shoulder", and the axes of the last three joints meet in  $p_w$ , since the manipulator wrist is spherical. In analogy with the human arm, the central joint represents therefore the elbow mobility, and the kinematic (or intrinsic) redundancy of the manipulator can be addressed by evaluating the elbow center point  $p_e$  when  $p_s$  and  $p_w$  are fixed, leading to the so-called elbow angle or swivel angle approach. Since the spherical wrist joint is accountable for the end-effector frame  $\{7\}$  orientation, its position is instead ruled by the first four revolute joints, that allow the rotation of the elbow point  $p_e$  along a circle  $S$  of radius  $R$  (Fig. 5). It is worth noticing that the swivel circle  $S$  can be seen as the intersection of two cones that have  $p_e$  and  $p_w$  as their respective vertices, while the apothems depend on the geometric parameters  $D_1, D_2, D_3, D_4$  and  $e_2$ . In detail, while the apothem  $L$  of the cone with vertex in  $p_s$  is  $U = D_1 + D_2$ , the geometric offset  $e_2$  causes a misalignment in the serial kinematic chain, thus the apothem of the second cone does not coincide with the sum  $D_3 + D_4$ . By observing Fig. 6, it is quite straightforward that, by defining:

$$L = D_3 + D_4 \tag{4}$$



**Figure 6.** Representation of the characteristic parameters of the elbow angle IK approach .

the apothem of the cone with  $p_w$  as vertex can be computed as:

$$L_{\perp} = \sqrt{L^2 + e_2^2} \quad (5)$$

Thus, to make use of the elbow angle approach, the position of the wrist centre  $p_w$  can be derived as follows:

$$\mathbf{p}_w = \mathbf{p}_7^b - (D_6 + D_7) [r_{713}^b, r_{723}^b, r_{733}^b] \quad (6)$$

where  $\mathbf{p}_7^b$  is the given goal position of the end-effector frame with respect to  $\{b\}$  and  $[r_{713}^b, r_{723}^b, r_{733}^b]$  is the third unit vector of the reference frame. To compute the  $\{\hat{\mathbf{u}}, \hat{\mathbf{v}}, \hat{\mathbf{n}}\}$  triplet, the following approach can be used:

$$\hat{\mathbf{n}} = \frac{\mathbf{p}_w - \mathbf{p}_s}{|\mathbf{p}_w - \mathbf{p}_s|} \quad (7)$$

$$\hat{\mathbf{u}} = \frac{\hat{\mathbf{u}} - (\hat{\mathbf{u}} \cdot \hat{\mathbf{n}}) \hat{\mathbf{n}}}{|\hat{\mathbf{u}} - (\hat{\mathbf{u}} \cdot \hat{\mathbf{n}}) \hat{\mathbf{n}}|} \quad (8)$$

$$\hat{\mathbf{v}} = \hat{\mathbf{n}} \wedge \hat{\mathbf{u}} \quad (9)$$

where  $\hat{\mathbf{a}}$  is an arbitrarily defined unit vector and the underlined notation defines the position vector with respect to the  $\{b\}$  frame. It is worth underlining that  $\hat{\mathbf{n}}$  is by definition perpendicular to the swivel circle  $S$ . The  $\gamma$  angle, i.e. the semi-vertical angle of the cone with  $p_s$  as the vertex, can be computed using the law of cosines with the two triangles  $p_s, p_c, p_e$  and  $p_c, p_e$  and  $p_w$ :

$$\gamma = \cos^{-1} \left( \frac{U^2 + |\mathbf{p}_w - \mathbf{p}_s|^2 - L_{\perp}^2}{2 |\mathbf{p}_w - \mathbf{p}_s| U} \right) \quad (10)$$

The position vector  $\mathbf{p}_c$  of the swivel circle centre and its radius  $R$  can be easily computed:

$$\mathbf{p}_c = \mathbf{p}_s + U \cos(\gamma) \hat{\mathbf{n}} \quad (11)$$

$$R = U \sin(\gamma) \quad (12)$$

The position vector of the elbow point  $\mathbf{p}_e$  can now be defined as a function of the derived parameters and the swivel angle  $\phi$ , i.e. the angle formed between  $\hat{\mathbf{u}}$  and  $\mathbf{p}_e - \mathbf{p}_c$ , and thus it represents the intrinsic kinematic redundancy of the manipulator:

$$\mathbf{p}_e = \mathbf{p}_c + R (\cos(\phi) \hat{\mathbf{u}} + \sin(\phi) \hat{\mathbf{v}}) \quad (13)$$

## 2.2. Joint Angles Computation

Once the position vector  $\mathbf{p}_e$  has been computed as function of the swivel angle parameter  $\phi$ , its value can be used to compute the first two joint angles  $q_1$  and  $q_2$ . Indeed, by referring to the homogeneous transformation matrix  $\mathbf{T}_3^b$ , that describes the position and orientation of the reference frame  $\{3\}$  with respect to  $\{b\}$ , the following applies:

$$\mathbf{T}_3^b = \mathbf{T}_0^b \mathbf{T}_1^0 \mathbf{T}_2^1 \mathbf{T}_3^2 = \begin{bmatrix} s_1 s_3 + c_1 c_2 c_3 & c_1 s_2 & c_3 s_1 - c_1 c_2 s_3 & c_1 s_2 (D_2 + D_3) \\ c_1 s_3 - c_2 c_3 s_1 & -s_1 s_2 & c_1 c_3 + c_2 s_1 s_3 & -s_1 s_2 (D_2 + D_3) \\ c_3 s_2 & -c_2 & -s_2 s_3 & D_1 - c_2 (D_2 + D_3) \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad (14)$$



Since  $\mathbf{p}_e = \mathbf{p}_3^b$ , it is possible to compute the first two joint angles as follows:

$$q_1 = -\tan^{-1} \left( \frac{p_{3y}^b}{p_{3x}^b} \right) \quad (15)$$

$$q_2 = \tan^{-1} \left( \frac{\sqrt{p_{3x}^b{}^2 + p_{3y}^b{}^2}}{D_1 - p_{3z}^b} \right) \quad (16)$$

The third joint value  $q_3$  can be calculated by imposing the following identity:

$$\mathbf{p}_e - e_2 [r_{313}^b, r_{323}^b, r_{333}^b] = \mathbf{p}_w + L [r_{413}^b, r_{423}^b, r_{433}^b] \quad (17)$$

where  $[r_{313}^b, r_{323}^b, r_{333}^b]$  and  $[r_{413}^b, r_{423}^b, r_{433}^b]$  are the third components of the reference frames  $\{3\}$  and  $\{4\}$ , that can be extracted from their respective homogeneous transformation matrices:

$$[r_{313}^b, r_{323}^b, r_{333}^b] = [c_3 s_1 - c_1 c_2 s_3, c_1 c_3 + c_2 s_1 s_3, -s_2 s_3] \quad (18)$$

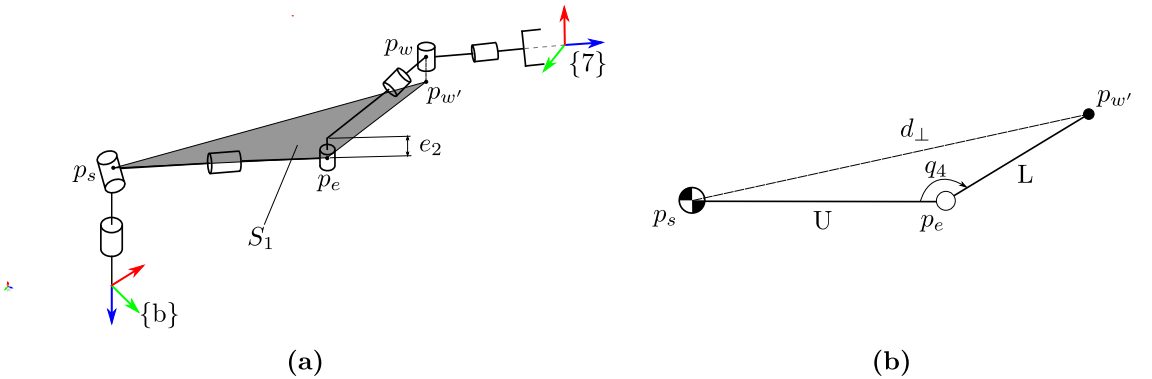
$$[r_{413}^b, r_{423}^b, r_{433}^b] = \begin{bmatrix} -s_4(c_1 c_2 c_3 + s_1 s_3) + c_4(c_1 s_2) \\ -s_4(s_1 c_2 c_3 + c_1 s_3) + c_4(-s_1 s_2) \\ -s_4 s_2 c_3 - c_4 c_2 \end{bmatrix} \quad (19)$$

Even though Eqn. 17 is function of  $q_3$  but also  $q_4$ , it can still be solved since  $q_4$ , i.e. the elbow angle, can be directly computed through geometric construction. As presented in Fig. 7 (a), the  $p_w$  point can be translated with a fixed off-set  $e_2$  along the  $e_2$  direction, thus obtaining the point  $p_{w'}$ . Looking at the  $\{p_s, p_e, p_{w'}\}$  triangle on the  $S_1$  plane, Fig. 7 (b), the value of the elbow joint  $q_4$  can be easily obtained through the cosine law:

$$q_4 = \cos^{-1} \left( \frac{L^2 + U^2 - d_{\perp}^2}{2UL} \right) \quad (20)$$

where

$$d_{\perp} = \sqrt{|\mathbf{p}_w - \mathbf{p}_s|^2 + e_2^2} \quad (21)$$



**Figure 7.** (a) Computation of the elbow angle  $q_4$  through geometric construction. (b) The wrist center point  $p_w$  is projected onto the  $S_1$  plane.

$q_3$  can now be computed as:

$$c_3 = - \frac{e_2 s_2 p_{w_x} - e_2 s_2 p_{e_x} + c_1 c_2 e_2 p_{e_z} - c_1 c_2 e_2 p_{w_z} +}{s_2 (s_1 L^2 s_4^2 + s_1 e_2^2)} \quad (22a)$$

$$\frac{+L s_1 s_4 p_{e_z} - L s_1 s_4 p_{w_z} + L c_1 c_2^2 c_4 e_2 + L c_1 c_4 e_2 s_2^2 + L^2 c_2 c_4 s_1 s_4}{s_2 (s_1 L^2 s_4^2 + s_1 e_2^2)}$$

$$s_3 = \frac{e_2 s_1 p_{w_z} - e_2 s_1 p_{e_z} - L s_2 s_4 p_{e_x} + L s_2 s_4 p_{w_x} + L^2 c_1 c_2^2 c_4 s_4 +}{s_2 (s_1 L^2 s_4^2 + s_1 e_2^2)} \quad (22b)$$

$$\frac{+L^2 c_1 c_4 s_2^2 s_4 - L c_2 c_4 e_2 s_1 + L c_1 c_2 s_4 p_{e_z} - L c_1 c_2 s_4 p_{w_z}}{s_2 (s_1 L^2 s_4^2 + s_1 e_2^2)}$$

$$q_3 = \tan^{-1} \left( \frac{s_3}{c_3} \right) \quad (22c)$$

Regarding the values of the last three joint angles, the standard inverse kinematics procedure for spherical joint was applied, thus by evaluating the  $T_7^4$  homogeneous transformation matrix:

$$\mathbf{T}_7^4 = \mathbf{T}_b^4 \mathbf{T}_7^b \quad (23)$$

where:

$$\mathbf{T}_b^4 = \begin{bmatrix} \mathbf{R}_4^{bT} & -\mathbf{R}_4^{bT} \mathbf{p}_4^b \\ [0, 0, 0] & 1 \end{bmatrix} \quad (24)$$

Finally,  $q_5, q_6, q_7$  can be computed:

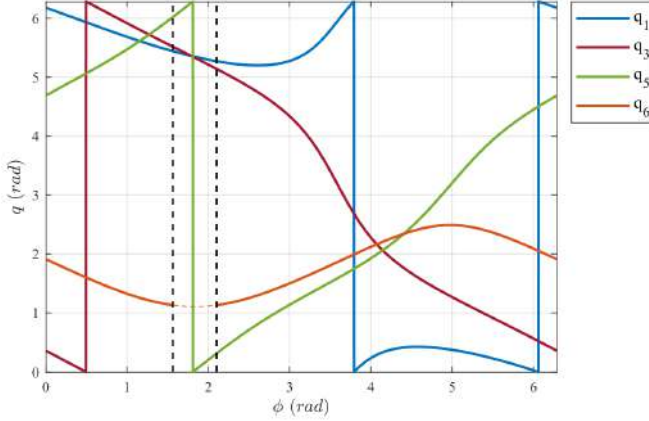
$$q_5 = \tan^{-1} \left( \frac{r_{7\ 23}^4}{r_{7\ 13}^4} \right) \quad (25)$$

$$q_7 = \tan^{-1} \left( \frac{-r_{7\ 32}^4}{-r_{7\ 31}^4} \right) \quad (26)$$

$$q_6 = \tan^{-1} \left( \frac{r_{7\ 32}^4}{\sin(q_7)} \right) \quad (27)$$

### 2.3. Redundancy Evaluation

Since the swivel angle approach is a closed-form solution of the Inverse Kinematics problem for the Jaco2 manipulator, it can be used to extract the entire range of possible postures that corresponds to a given goal pose  $\mathbf{T}_7^b$ . Indeed, it is worth noticing the manipulator joint space has size  $m = 7$  while its workspace has size  $n = 6$ , thus the arm has  $m - n = 1$  degree of kinematic redundancy. Typically, this feature is exploited to augment the mobility of the manipulator, e.g. for collision avoidance algorithms. The typical behavior of the joint angle values as a function of the swivel angle  $\phi$  is presented in Fig. 8, where the curves are presented in the range  $[0, 2\pi]$ . Focusing on joint 6 curve, the whole set of possible solutions is restricted due to the physical joint limit constraints, that impose a minimum



**Figure 8.** Behavior of the values of the joint angles in the interval  $[0, 2\pi]$  as a function of the swivel angle  $\phi$  for a given pose. The black vertical dotted lines demarcate the not allowed solutions due to joint 6 limit constraints. .

admitted value on  $q_6$ . In general, the Kinova Jaco2 manipulator presents physical joint limits in joint 2,4,6, as presented in Tab. 3. Thus, it seems evident that the closed-form solution can also be exploited to avoid those solutions that are outside or even close to the admitted joint values boundary. To this aim, Vahrenkamp et al. [25] presented a novel approach in which an extended manipulability index, that takes into account closeness to joint limits and possible obstacles, is presented. Starting from Togai's

**Table 3.** Physical joint limit values for Kinova Jaco2 manipulator.

q	$q_{min}$ (deg)	$q_{max}$ (deg)
2	47	313
4	30	330
6	65	295

definition of manipulability index [26], that estimates the posture closeness to kinematic singularities:

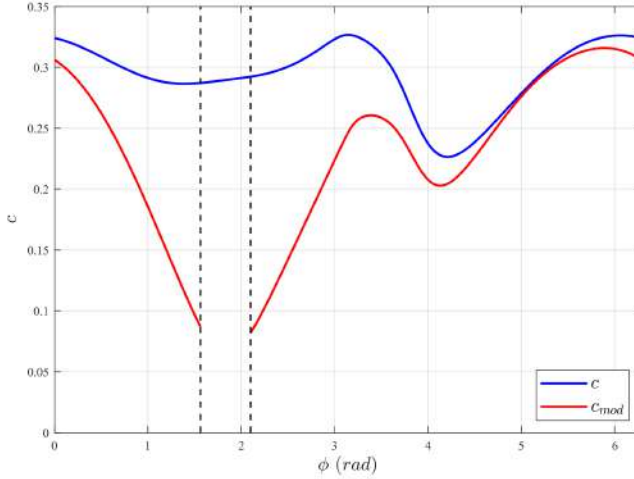
$$c(\mathbf{J}) = \frac{1}{\text{cond}(\mathbf{J})} = \frac{\lambda_{min}}{\lambda_{max}} \quad (28)$$

where  $\mathbf{J} = \mathbf{J}(q_1, \dots, q_7)$  is the arm Jacobian matrix and  $\lambda_{min,max}$  are its minimum and maximum eigenvalues, in [25, 27] the following penalization function  $p_j$  for the  $j$ -th column of the Jacobian is presented:

$$p_j = \frac{1}{\sqrt{1 + |\nabla h(q_j)_j|}} \quad (29a)$$

$$\nabla h(q_j)_j = \frac{(\theta_{j,max} - \theta_{j,min})^2 (2\theta_j - \theta_{j,max} - \theta_{j,min})^2}{\xi (\theta_{j,max} - \theta_j)^2 (\theta_j - \theta_{j,min})^2} \quad (29b)$$

where  $q_{j,max}$ ,  $q_{j,min}$  were presented in Tab. 3 and  $\xi$  is an arbitrarily defined parameter that accentuates or relaxes the  $p_j$  function. In detail, it is easy to check that larger values of  $\xi$  increase the area where  $p_j$  has a value close to one. In this paper, the value of  $\xi = 4$ , as used in [25], was selected. The manipulability



**Figure 9.** Manipulability index as a function of the elbow angle for a defined goal pose. Vertical dashed lines denote the area not allowed due to joint limit constraints.

$c$  can then be modified into  $c_{mod}$  evaluating the condition number of the modified Jacobian, where the  $j$ -th column of the matrix is multiplied with the corresponding penalization factor  $p_j$ :

$$J_{mod_{i,j}} = J_{i,j} p_j \quad i = 1, \dots, 6, \quad j = 1, \dots, 7 \quad (30)$$

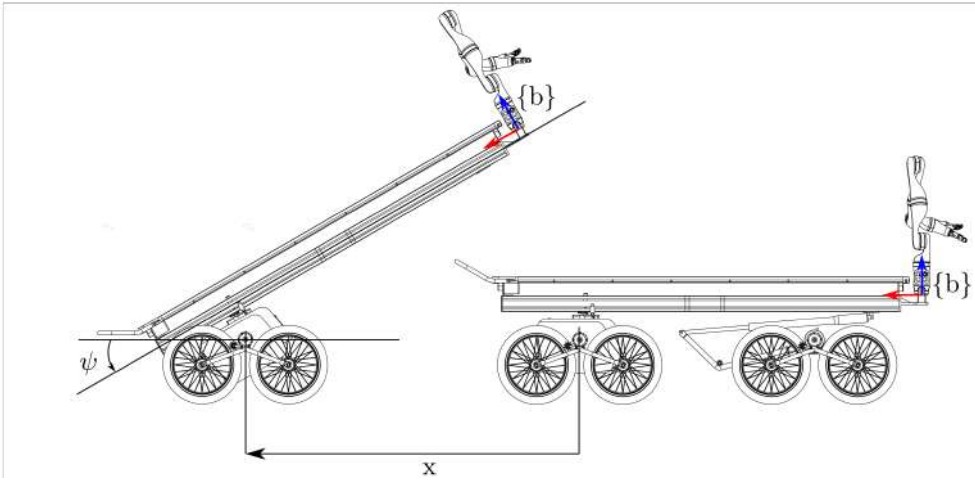
The behavior of  $c$  and  $c_{mod}$  is presented in Fig. 9, where the closeness to the joint 6 physical limit leads to a significant penalization of  $c$ . The implementation of the manipulability index  $c_{mod}$  therefore establishes a criterion to evaluate the "best" posture that corresponds to a defined goal pose, thus allowing to extract a solution from the Inverse Kinematics set presented in Fig. 8.

### 3. System Implementation

To properly navigate and collect crops, the manipulator mobility can be augmented thanks to Agri.Q kinematics, i.e. the mobile base mobility, thus making the whole system a mobile manipulator. Even though the Agri.Q mobility is quite complex, due to the articulated architecture and the potential eight drive wheels, as presented by the authors in [28, 29], its kinematics is here simplified for crop collecting purposes, that can be classified as pick-and-place tasks.

When navigating inside vineyard rows, the two degrees of freedom that can be used to augment the manipulator mobility are the linear displacement  $x$  and the pitch angle  $\psi$ , as presented in Fig. 10. While is quite straightforward the impact of  $x$ , it is worth underlining the pitch angle has the double effect of moving the arm upwards but also rotate it.

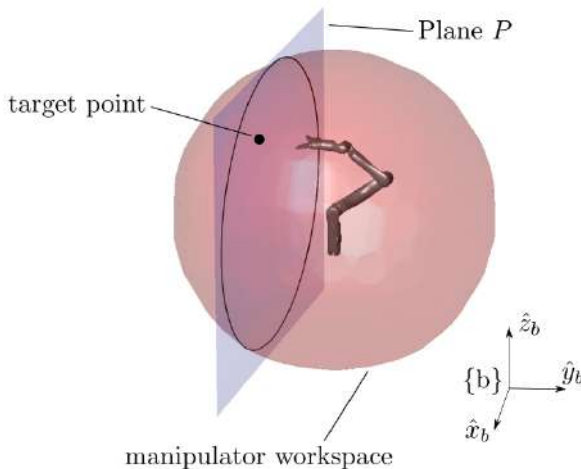
Even though several motion planning algorithms extend the Jacobian of the arm with the base one, thus moving towards the simultaneous motion of the base and the arm [30, 31], the motion planning is here addressed with the decoupling of the mobile base motion and the manipulator one, due to the significant difference between the position accuracy of the two systems. According to this approach, the linear displacement  $x$  and the pitch angle  $\psi$  are used to position the manipulator near the goal grape peduncle, then the arm is used to perform the pick and place task. Since the base mobility can't affect the displacement of a grape along the  $y$  axis of  $\{b\}$ , the base can translate and rotate the arm workspace



**Figure 10.** *Agri.Q base mobility simplification for crop sampling.*

(Fig. 11) to position the goal point inside the area where the manipulability index  $c_{mod}$  has the highest value [28]

It is worth underlining that, although it significantly simplifies the motion planning pipeline, the proposed approach can't be used for dynamic handling or grasping, since it is mainly based on a static representation of the environment and the Agri.Q itself. It is also assumed that the Agri.Q base can fulfill the arm positioning task with the theoretical absence of positioning errors, which is quite impossible to obtain in a real vineyard. This aspect is partially compensated since the target point is positioned inside the area where the Jaco2 manipulator presents the highest value of manipulability index. Thus, if the positioning error is restricted, the target point still lies into the best manipulability area. Nonetheless, the evaluation of the cited area depends on the pose of the target object, i.e. its position along the  $y$



**Figure 11.** *Use of the Agri.Q mobility to translate and rotate the arm workspace. Plane  $P$  is parallel to  $\langle \hat{x}_b, \hat{z}_b \rangle$  and its displacement along  $\hat{y}_b$  is established by the position of the target point.*

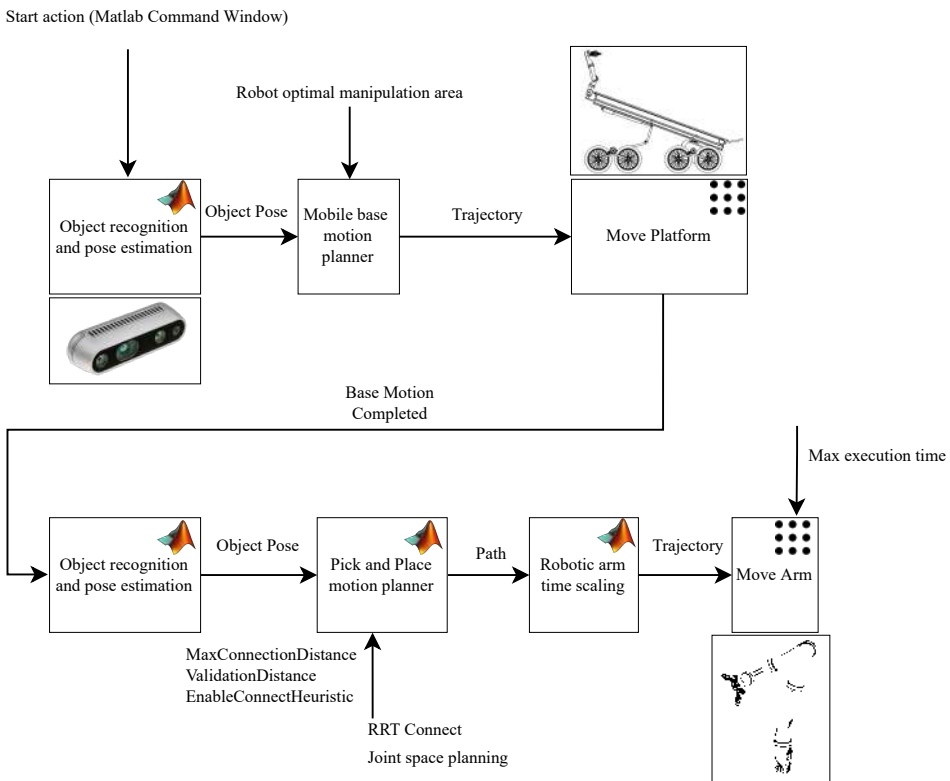
axis of  $\{b\}$  and its orientation, so it would be outperformed for each target grapevine, with a significant increment in the overall computational time.

### 3.1. Software Development

Within the present sub-section, a comprehensive description of the developed software architecture for the system control is presented.

Regarding the closed-form Inverse Kinematics method presented in Section 2, the developed algorithm is hereafter reported in pseudo-code form in Alg. 1. Since it can be potentially used for different research activities, the IK code is provided as open-source Matlab code by the authors<sup>1</sup>. The library was developed using the Matlab multidimensional matrices functionality, to optimize the computation time especially in case of redundancy evaluation, where a significant reduction of the computational cost from the average value of 0.65s to 0.15s was obtained<sup>2</sup>. The code was further optimized with the Matlab .mex code generation functionality, where the average computation time of 0.09s was achieved<sup>3</sup>. The code uses the *Solveq* sub-function that performs the joint angles computation according to Subsection 2.2.

Regarding the Agri.Q system implementation, in Fig. 12 a representation of a first and simplified system software architecture is presented, where the interaction between Matlab and ROS environment was



**Figure 12.** Software system architecture. The Matlab and ROS logo are used to indicate the selected environment for a specific block.

<sup>1</sup><https://github.com/giocolucci/Jaco2SwivelIK>

<sup>2</sup>Tested on tested on a Dell XPS machine, with Ubuntu 20.04 LTS, and an Intel i7-10510U@1.80 GHz processor. The computation time is evaluated as the average of 20 tests.

<sup>3</sup>If requested, the .mex file can be provided by the authors.

**Algorithm 1** Inverse Kinematics elbow angle algorithm

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```

angle ← [α, β, γ]                                ▶ The tool center point represented as XYZ Euler angles

 $\mathbf{T}_{7goal}^b \leftarrow eul2tform(angle, 'XYZ')$     ▶ The goal  $\mathbf{T}_7^b$  4x4 homogeneous transformation matrix

 $\mathbf{T}_{7goal}^b(1 : 3, 4) \leftarrow [p_x, p_y, p_z]$         ▶ The tool center point goal position is inserted into  $\mathbf{T}_7^b$ 

SwivelIKOption ← 'Fixed' or 'Optimized' or 'All';    ▶ To identify the IK algorithm output

if SwivelIKOption = 'Fixed' then                  ▶ The solution at a given elbow angle  $\phi$  is computed
     $\phi \leftarrow \phi_{goal}$ 
     $\mathbf{p}_e \leftarrow \mathbf{p}_c + R \cos(\phi) \hat{\mathbf{u}} + R \sin(\phi) \hat{\mathbf{v}}$ ;    ▶  $\mathbf{p}_e$  is a 1x3 vector
     $\mathbf{q} \leftarrow Solveq(\mathbf{p}_e, \mathbf{T}_{7goal}^b)$ ;    ▶ The output is a 7x1 joint angles vector

else if SwivelIKOption = 'Optimize' or 'All' then
     $\phi \leftarrow linspace(0, 2\pi, n_{points})$ ; ▶ The continuous set of solutions is discretized into  $n_{points}$  values
     $\mathbf{p}_e \leftarrow \mathbf{p}_c + R \cos(\phi) \hat{\mathbf{u}} + R \sin(\phi) \hat{\mathbf{v}}$ ;    ▶  $\mathbf{p}_e$  is a  $n_{points} \times 3$  matrix
     $\mathbf{q} \leftarrow Solveq(\mathbf{p}_e, \mathbf{T}_{7goal}^b)$ ;
    Compute  $\mathbf{J}_{mod}$  taking into account the penalization factors

    for counter = 1 :  $n_{points}$  do
        if the posture is feasible then
             $c_{mod}(counter) \leftarrow \frac{1}{cond(\mathbf{J}_{mod}(:, :, counter))}$ ;
        else if the physical joint limit constraints don't allow the posture then
             $c_{mod}(counter) \leftarrow 0$ ;
        end if
    end for

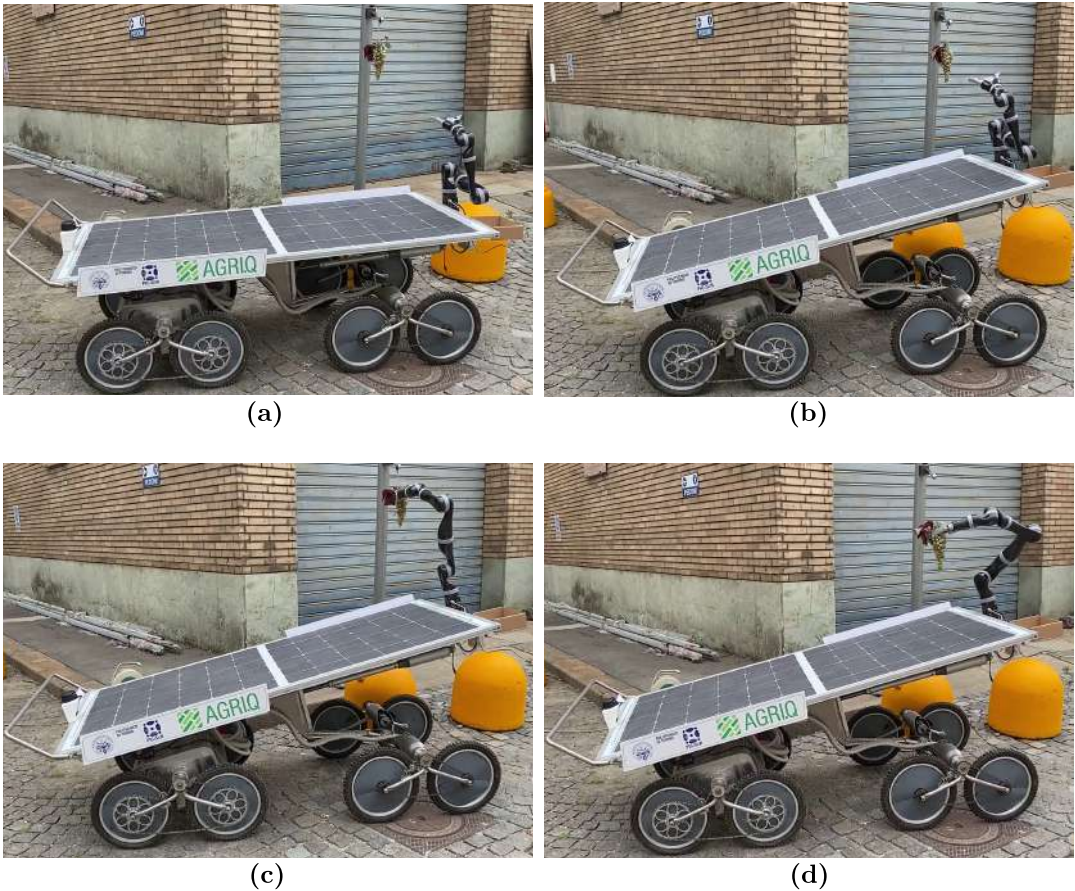
    if SwivelIKOption = 'Optimize' then ▶ The solution at which corresponds the biggest  $c_{mod}$  value is computed
         $[c_{modmax}, index] \leftarrow max(c_{mod})$ ;
         $\mathbf{q} \leftarrow q(index, :)$ ;    ▶ The output is a 7x1 joint angles vector

        else if SwivelIKOption = 'All' then ▶ The whole set of solution, discretized according to the  $\phi$  elbow angle vector definition, is computed
             $\mathbf{q} \leftarrow \mathbf{q}$     ▶ The output is a  $7 \times n_{points}$  joint angles matrix
             $\mathbf{c}_{mod} \leftarrow \mathbf{c}_{mod}$ 
        end if
    end if

```

---

done through the Matlab ROS Toolbox [32]. A camera block will be implemented for object recognition and pose estimation of the target, i.e. the grape peduncle, both before and after the mobile base motion to avoid possible positioning errors of the mobile base, while the manipulator path planning pipeline involves the use of standard sample-based joint space path planning like RRT connect [33]. Except for the camera block, the presented pipeline was used to perform a first laboratory experimental validation, as shown in Fig. 13. Agri.Q was able to pick-and-place a grapevine sample.



**Figure 13.** *Agri.Q* prototype first experimental tests. (a) Rest configuration, (b) goal peduncle approaching, (c) peduncle picking, (d) place of the peduncle .

#### 4. Conclusions

Within this paper, the development and implementation of the motion planning pipeline for *Agri.Q* mobile manipulator for Precision Agriculture was presented. The algorithm had integrated the resolution of the arm kinematic redundancy through the closed-form elbow angle formulation, that is provided by the author as an open source Matlab code. The whole system motion pipeline was instead addressed using a motion decoupling approach, where the mobile base translate and rotate the manipulator workspace to position the goal point inside the area with the highest value of the manipulability index. The whole architecture was implemented on the real prototype and validated through first tests in laboratory environment. As further developments, a perception block will be developed to both recognize the target peduncle and also provide a feedback after the mobile base motion to correct possible positioning errors.

**Author Contributions.** GC, AB, LT and GQ conceived and designed the study. GC, LT and AB carried out the software implementation. GQ validated the study. GC and AB prepared the draft of the paper. GQ, LT and LB edited and reviewed the draft.

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**Ethical Approval.** Not applicable.

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