On solving the direct kinematics problem for parallel robots

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Abstract: In this paper, we propose efficient methods for solving the direct kinematics problem (DKP) for parallel manipulators. By solving, we mean computing all the real solutions in a certified way, without any assumption on the manipulator. For example, the precision and the real character of the solutions are guaranteed. The proposed algorithms use as black boxes several recent algorithms from computer algebra such as F4/F5 for computing Gröbner bases, recent methods for isolating the real roots by mean of intervals with rational bounds which are mixed with strategies coming from the interval arithmetics field. The resulting solutions are efficient for such kind of output since the running time never exceeds few seconds (in fact about 1 second for non extreme but general examples) thanks to the algorithms but also to a well adapted translation of DKP problem into algebraic equations. As an example we give a new computational proof of the existence of a parallel robot with 40 real roots.

Key-words: Direct Kinematics Problem, Parallel Robot, Polynomial System Solving, Computer Algebra, Real Roots.

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Sur la résolution du problème géométrique direct pour les robots parallèles

Résumé : Nous proposons une méthode efficace pour résoudre le problème géométrique direct (DKP) pour les robots parallèles. Le terme résoudre veut dire ici calculer toutes les solutions réelles de façon certifiée sans aucune supposition sur le robot. Par exemple, la précision ou le caractère réel des solutions sont garantis. Les méthodes proposées utilisent différents algorithmes de calcul formel récents tels que F4/F5 pour le calcul de bases de Gröbner ou pour l’isolation des racines réelles de polynômes par des intervalles à bornes rationnelles qui sont associés avec d’autres stratégies utilisant de l’arithmétique par intervalles. L’algorithme final est efficace puisque les temps de calcul sont voisins de la seconde sur des exemples généraux, grâce, en particulier, à une formulation adaptée du DKP. Comme exemple, nous donnons une preuve calculatoire de l’existence d’un robot parallèle avec 40 positions.

Mots-clés : Problème géométrique direct, Robot parallèle, Résolution de systèmes polynomiaux, Calcul Formel, Racines réelles
On solving the direct kinematics problem for parallel robots

Contents

1 Introduction and notations .................................................. 3

2 Gröbner based Real Solving .............................................. 5
  2.1 Notations and basic definitions .................................... 5
  2.2 Gröbner bases ...................................................... 5
  2.3 Zero-dimensional systems ......................................... 8
  2.4 The Rational Univariate Representation .......................... 9
  2.5 From formal to numerical solutions .............................. 10

3 Algebraic modelisation ..................................................... 11
  3.1 Standard algebraic equations ...................................... 11
  3.2 Quaternions ......................................................... 13
    3.2.1 The generic case ............................................. 13
    3.2.2 The non-generics case ...................................... 14
    3.2.3 Recovering the real solutions of the DKP ................. 14

4 Experiments ....................................................................... 15

5 Conclusion ........................................................................ 19

1 Introduction and notations

Most industrial robots have a serial mechanical architecture i.e. starting from the fixed base each link is connected through an actuated joint to the next link in the chain. Such structure provides large workspace but is not appropriate in terms of accuracy and payload. For applications requiring a good positioning accuracy and/or large payload it has been suggested the use of closed-loop mechanical structure in which more than one kinematic chains connect the ground and the end-effector. A typical example of such structure is the Gough platform [12], figure 1.

In this mechanism, the end-effector is connected to the ground by 6 articulated legs. The extremities $A_i, B_i$ of the legs are connected to the ground and to the end-effector by spherical joints, allowing free rotations around the joint centers. In each leg is a linear actuator that allows to change the leg lengths and a sensor that allow to measure the leg length. By changing the 6 leg lengths it is possible to control the position/orientation (called a pose) of the end-effector with respect to the ground. For controlling such robot it is essential to solve the following two kinematic problems: the inverse kinematic problem (IKP) and the direct kinematic problem (DKP). For the IKP the position/orientation of the platform, defined by a set of parameters X is known and we must determine the corresponding leg lengths $\rho$. A simple analysis allows to obtain an analytical relation $\rho = F(X)$. More generally the IKP is to determine the actuated joint variables as a function of the end-effector pose parameters. The DKP is the opposite problem: being given the leg lengths it is necessary to determine
Figure 1: The Gough platform

the current position/orientation of the platform. Determining the position/orientation is essential for the use of the robot as any control law will use this information. Two types of DKP can be considered:

1. off-line: when starting the robot it is necessary to determine the start pose of the robot with as only available information the leg lengths as measured by the sensors. Computation time is not critical

2. real-time: it will be used for control purpose while the robot is moving and as soon as measurements are available. But here we may assume that the current pose is very close to the one obtained at the previous calculation.

In this paper we address mainly the off-line DKP (real-time DKP may be solved in a safe way by other methods [18]) but our purpose is to design a solving algorithm that is numerically robust, provide certified solutions and is reasonably fast for off-line studies. By certified solutions we mean that no spurious one will be found (in particular we compute only the real solutions) and that we will be able to refine them up to any arbitrary accuracy. Certification is a key issue for the DKP as any error will leads to a failure in the control of the robot that may have catastrophic consequences. This rules out some solving methods such as elimination whose last numerical step cannot be certified. Indeed another use of the DKP is to simulate the real robot which differs from its theoretical model (e.g. manufacturing tolerances induce uncertainties in the location of the $A_i, B_i$ points and the control law will exhibit positioning errors when the robot follows a nominal trajectory).

Methods from computer algebra have already been used for several purpose in robotics, and especially for solving/studying the DKP. For example, in [13] or [29], a univariate polynomial with as roots as the DKP is explicitly computed. But one can notice that those computer algebra based methods are mainly designed for getting qualitative results (mainly the number of solutions). One goal of this article is to promote such methods by showing
that recent progress make now possible their use for solving the DKP efficiently (few seconds in the worth case) and for any robot (no assumption on the geometry).

In the first section of the present contribution, we propose some general strategies for computing efficiently the real roots of algebraic systems of equations through Gröbner bases.

In the second section, we show how to model the DKP in order to speed up significantly the computations using the algorithms of the first section.

The last section is devoted to experiments on various class of robots. We illustrate the power of the method by computing and comparing the positions of a theoretical planar robot (left-hand) with those of an effective one (the same after small perturbations), studying the behavior of the method with respect to the precision of the input data. As a second example, we prove algorithmically that Dietmaier’s robot [E] has 40 real roots: by using Computer Algebra methods we can certify that all the solutions are real numbers (and not only complex numbers with small imaginary parts).

## 2 Gröbner based Real Solving

This section briefly summarizes some basic results about Gröbner bases and their applications to solving zero-dimensional systems (systems with a finite number of complex roots). The reader interested in a complete introduction of Gröbner bases and the required background is referred to \[3,4].

### 2.1 Notations and basic definitions

Let denote by \( \mathbb{Q}[X_1, \ldots, X_n] \) the ring of polynomials with rational coefficients and unknowns \( X_1, \ldots, X_n \) and \( S = \{ P_1, \ldots, P_s \} \) any subset of \( \mathbb{Q}[X_1, \ldots, X_n] \). A point \( x \in \mathbb{C}^n \) is a zero of \( S \) if \( P_i(x) = 0 \quad \forall i = 1 \ldots s. \) The ideal \( \mathcal{I} = \langle P_1, \ldots, P_s \rangle \) generated by \( P_1, \ldots, P_s \) is the set of polynomials in \( \mathbb{Q}[X_1, \ldots, X_n] \) constituted by all the combinations \( \sum_{k=1}^{s} P_k U_k \) with \( U_k \in \mathbb{Q}[X_1, \ldots, X_n] \). Since every element of \( \mathcal{I} \) vanishes at each zero of \( S \), we denote by \( V_c(S) = V_c(\mathcal{I}) = \{ x \in \mathbb{C}^n \mid p(x) = 0 \quad \forall p \in \mathcal{I} \} \) (resp. \( V_\mathbb{R}(S) = V_\mathbb{R}(\mathcal{I}) = V_c(\mathcal{I}) \cap \mathbb{R}^n \)) the set of complex (resp. real) zeroes of \( S \).

### 2.2 Gröbner bases

A Gröbner basis of \( \mathcal{I} \) is a computable generator set of \( \mathcal{I} \) with good algorithmical properties (as described below) and defined with respect to a monomial ordering. In this paper, one will use the following orderings:

- **lexicographic order** : (Lex)

\[
X_1^{\alpha_1} \cdots X_n^{\alpha_n} <_{\text{Lex}} X_1^{\beta_1} \cdots X_n^{\beta_n} \iff \exists i_0 \leq n, \quad \left\{ \begin{array}{l}
\alpha_i = \beta_i, \quad \text{for } i = 1, \ldots, i_0 - 1, \\
\alpha_{i_0} < \beta_{i_0}
\end{array} \right.
\]

\[ (1) \]
• degree reverse lexicographic order \((DRL)\):

\[
X_1^{a_1} \cdots X_n^{a_n} <_{DRL} X_1^{b_1} \cdots X_n^{b_N} \iff X_0^{\sum_k a_k} X_1^{-a_1} \cdots X_n^{-a_n} <_{Lex} X_0^{\sum_k b_k} X_1^{-b_1} \cdots X_n^{-b_n}
\]  

\[(2)\]

Gröbner bases are mathematical objects with good algorithmical properties: the main ones are summarized in the following.

**Definition 1** For any \(n\)-uple \(\mu = (\mu_1, \ldots, \mu_n) \in \mathbb{N}^n\), let denote by \(X^\mu = X_1^{\mu_1} \cdots X_n^{\mu_n}\). If \(<\) is an admissible monomial ordering and \(P = \sum_{i=0}^r a_i X_i^{\nu_i}\) any polynomial in \(\mathbb{Q}[X_1, \ldots, X_n]\), we define: \(\text{LM}(P, <) = \max_{i=0, \ldots, r} \nu_i\) (leading monomial of \(P\) w.r.t. \(<\)).

**Definition 2** A set of polynomials \(G\) is a Gröbner basis of an ideal \(I\) wrt to a monomial ordering \(<\) if for all \(f \in I\) there exists \(g \in G\) such that \(\text{LM}(g)\) divides \(\text{LM}(f)\).

Given any admissible monomial ordering \(<\) one can easily extend the classical Euclidean division to reduce a polynomial \(p\) by another one or, more generally, by a set of polynomials \(F\), performing the reduction wrt each polynomial of \(F\) until getting an expression which can not be reduced anymore. We denote this function by \(\text{Reduce}(p, F, <)\) (reduction of the polynomial \(p\) wrt \(F\)). Unlike in the univariate case, the result of such a process is not canonical except if \(F = G\) is a Gröbner basis:

**Theorem 1** Let \(G\) be a Gröbner basis of an ideal \(I \subset \mathbb{Q}[X_1, \ldots, X_n]\) for a fixed ordering \(<\).

(i) a polynomial \(p \in \mathbb{Q}[X_1, \ldots, X_n]\) belongs to \(I\) if and only if \(\text{Reduce}(p, G, <) = 0\),

(ii) \(\text{Reduce}(p, G, <)\) does not depend on the order of the polynomials in the list \(G\), thus, this is a canonical reduced expression modulus \(I\).

Gröbner bases are computable objects. The historical method for computing them is Buchberger’s algorithm [13]. It has several variants and it is implemented in most general computer algebra systems like Maple or Mathematica. Recently, more efficient algorithms have been proposed to compute Gröbner bases:

• the \(F_4\) algorithm [23] is based on the intensive use of linear algebra methods: in short, the arbitrary choices (\(S\)-polynomials) are left to computational strategies related to classical linear algebra problems (mainly the computation of row echelon form), the polynomials being represented as vectors in the monomial basis.

• In [11] a new criterion (the \(F_5\) criterion) for detecting useless computations (most of the \(S\)-polynomials to be computed using the classical Buchberger algorithm reduce to 0) has been given; under some regularity conditions on the system, it is proved that the algorithm do never perform useless computations. A new algorithm named \(F_5\) has been built using these two ideas: the \(F_5\) algorithm constructs incrementally the following matrices in degree \(d\):
where the indices of the columns are monomials sorted for the admissible ordering < and the rows are product of some polynomials $f_i$ by some monomials $t_j$ such that $\deg(t_j f_i) \leq d$. For a regular system the matrices, $A_d$ are full rank. In a second step, row echelon forms of the matrices are computed:

$$A'_d = \begin{bmatrix} t_1 f_1 & m_1 & m_2 & m_3 & \ldots \\ t_2 f_2 & 1 & 0 & 0 & \ldots \\ t_3 f_3 & 0 & 1 & 0 & \ldots \\ \vdots & \vdots & \vdots & \vdots & \ddots & \ldots \\ \end{bmatrix}$$

Even if $F_5$ still computes the same mathematical object (a Gröbner basis), the gap with existing other algorithms is consequent. In particular, due to the range of examples which become computable, Gröbner basis can be considered as a reasonable computable object in large applications. Important parameters to evaluate the complexity of Gröbner bases with the $F_5$ are the $D$ the maximal degree $d$ occuring in the computation and the size of the matrix $A_d$. The overall cost is thus dominated by $(\# A_d)^3$.

We pay a particular attention to Gröbner bases computed for elimination orderings since they provide a way to simplify the system (an equivalent system with a structured shape). For example, a lexicographic Gröbner basis of zero dimensional system (when the number of complex solutions is finite) has always the following shape (if we suppose that $X_1 < X_2 \ldots < X_n$):

$$\begin{align*}
  f(X_1) &= 0 \\
  f_2(X_1, X_2) &= 0 \\
  \vdots \\
  f_k(X_1, X_2) &= 0 \\
  f_{k+1}(X_1, X_2, X_3) &= 0 \\
  \vdots \\
  f_{k+1+n-1}(X_1, \ldots, X_n) &= 0 \\
  \vdots \\
  f_k(X_1, \ldots, X_n) &= 0 \\
\end{align*}$$

(depending on some geometrical properties of the solutions, some of the polynomials may be identically null). A well known property is that the zeros of the smallest (w.r.t. $<$) non null polynomial define the Zariski closure (classical closure in the case of complex solutions) of the projection on the coordinate's space associated with the smallest variables.
More generally, an admissible ordering $<$ on the monomials depending on variables $[X_1, \ldots, X_n]$ is an ordering which eliminates $X_{d+1}, \ldots, X_n$ if $X_i < X_j \forall i = 1 \ldots d, j = d + 1 \ldots n$. The lexicographic ordering is a particular elimination ordering. Given two monomial orderings $<_1$ (w.r.t. the variables $X_1, \ldots, X_d$) and $<_2$ (w.r.t. the variables $X_{d+1}, \ldots, X_n$) one can define an ordering which eliminates $X_{d+1}, \ldots, X_n$ by setting the so called block ordering $<_{1,2}$ as follows: given two monomials $m$ and $m'$, $m <_{1,2} m'$ if and only if $m|_{X_1=1, \ldots, X_n=1} <_{2} m'|_{X_1=1, \ldots, X_n=1}$ or $(m|_{X_1=1, \ldots, X_n=1} = m'|_{X_1=1, \ldots, X_n=1} \text{ and } m <_{1} m')$.

### 2.3 Zero-dimensional systems

Zero-dimensional systems are polynomial systems with a finite number of complex solutions. This specific case is fundamental for many engineering applications including parallel manipulator DKPs which are known to have, at most, 40 complex solutions in non degenerated cases [16,19,21]. The following known theorem shows that we can detect easily that a system is zero dimensional or not by computing a Gröbner base for any monomial ordering:

**Theorem 2** Let $G = \{g_1, \ldots, g_r\}$ be a Gröbner basis for any ordering $<$ of any system $S = \{P_1, \ldots, P_s\} \in \mathbb{Q}[X_1, \ldots, X_n]^s$. The two following properties are equivalent:

- For all index $i$, $i = 1 \ldots n$, there exists a polynomial $g_j \in G$ and a positive integer $n_j$ such that $X_i^{n_j} = \text{LM}(g_j, <)$;
- The system $\{P_1 = 0, \ldots, P_s = 0\}$ has a finite number of solutions in $\mathbb{C}^n$.

If $S$ is zero-dimensional, then, according to theorem 2, only a finite number of monomials $m \in \mathbb{Q}[X_1, \ldots, X_n]$ are not reducible modulo $G$, meaning that $\text{Reduce}(m, G, <) = m$. Mathematically, a system is zero-dimensional if and only if $\mathbb{Q}[X_1, \ldots, X_n]/I$ is a $\mathbb{Q}$-vector space of finite dimension, $I$ being the ideal generated by the system. This vector space can fully be characterized when knowing a Gröbner basis:

**Theorem 3** Let $S = \{p_1, \ldots, p_s\}$ be a set of polynomials with $p_i \in \mathbb{Q}[X_1, \ldots, X_n]$, $\forall i = 1 \ldots s$, and suppose that $G$ is a Gröbner basis of $I = \langle S \rangle$ with respect to any monomial ordering $<$. Then:

- $\mathbb{Q}[X_1, \ldots, X_n]/I = \{\text{Reduce}(p, G, <) \mid p \in I\}$ is a vector space of finite dimension;
- $B = \{t = \sum_{i=1}^n e_i x_i^{e_i} \mid (e_1, \ldots, e_n) \in \mathbb{N}^n \text{ and } \text{Reduce}(t, G, <) = t\} = \{w_1, \ldots, w_D\}$ is a basis of $\mathbb{Q}[X_1, \ldots, X_n]/I$ as a $\mathbb{Q}$-vector space;
- $D = 2^{|B|}$ is exactly the number of elements of complex zeroes of the system $\{P = 0, \forall P \in S\}$ counted with multiplicities.

Thus, in the zero-dimensional case, once a Gröbner basis is known, a basis of $\mathbb{Q}[X_1, \ldots, X_n]/I$ can easily be computed (Theorem 3) so that linear algebra methods can be applied for doing several computations and getting informations about the roots.
For any polynomial \( q \in \mathbb{Q}[X_1, \ldots, X_n] \) the decomposition \( \overline{q} = \text{Reduce}(q, G, <) = \sum_{i=1}^{D} a_i w_i \) is unique (Theorem 1) and we denote by \( \overline{q} = [a_1, \ldots, a_D] \) the representation of \( \overline{q} \) in the basis \( B \). For example, the matrix w.r.t. \( B \) of the linear map

\[
m_{\overline{q}} : \begin{pmatrix} \mathbb{Q}[X_1, \ldots, X_n]/I \\ \overline{q} \end{pmatrix} \rightarrow \mathbb{Q}[X_1, \ldots, X_n]/I
\]

can explicitly be computed (its columns are the vectors \( \overline{q}w_i \)) and one can then apply the following well-known theorem:

**Theorem 4 (Stickelberger)** The eigenvalues of \( m_q \) are exactly the \( q(\alpha) \) where \( \alpha \in V_C(S) \).

According to Theorem 4 the \( i \)-th coordinate of all \( \alpha \in V_C(S) \) can be obtained from \( M_{X_i} \) eigenvalues but the issue of finding all the coordinates of all the \( \alpha \in V_C(S) \) from \( M_{X_1}, \ldots, M_{X_n} \) eigenvalues is not explicit nor straightforward (see 1 for example). Note also that some authors propose algorithms to compute numerically the matrices \( M_{X_1}, \ldots, M_{X_n} \) without computing Gröbner bases (see 20). Up to our experiments, such computations are not numerically stable for general manipulators and it may be preferable to compute, for example, the characteristic polynomial of the matrix \( M_{X_i} \) and then isolate its real roots to obtain all the possible \( i \)-th coordinates of the solutions: we prefer to follow with exact computations a little bit more, providing exact formulas as explained in the next section.

### 2.4 The Rational Univariate Representation

The Rational Univariate Representation [26] is, with the end-user point of view, a simple way for representing symbolically the roots of a zero-dimensional system without loosing information (multiplicities or real roots) since one can get all the information on the roots of the system by solving exclusively univariate polynomials.

Given a zero-dimensional system \( I = \langle p_1, \ldots, p_s \rangle \) where the \( p_i \in \mathbb{Q}[X_1, \ldots, X_n] \), a Rational Univariate Representation of \( V(I) \) has the following shape: \( f_i(T) = 0, X_1 = \frac{g_{1,i}(T)}{g_{n,i}(T)}, \ldots, X_n = \frac{g_{n,i}(T)}{g_{n,i}(T)} \) where \( f_i, g_{1,i}, g_{2,i}, \ldots, g_{n,i} \in \mathbb{Q}[T] \) (\( T \) is a new variable). It is uniquely defined w.r.t. a given polynomial \( t \) which separates \( V(I) \) (injective on \( V(I) \)), the polynomial \( f_i \) being necessarily the characteristic polynomial of \( m_t \) (see above section) in \( \mathbb{Q}[X_1, \ldots, X_n]/I \) [26]. The RUR defines a bijection between the roots of \( F \) and those of \( f_i \) preserving the multiplicities and the real roots:

\[
\begin{align*}
V(S)(\cap \mathbb{R}) \\
\text{\alpha = (\alpha_1, \ldots, \alpha_n)} \\
(X_1(\alpha) = \frac{g_{1,i}(t(\alpha))}{g_{1,i}(t(\alpha))}, \ldots, X_n(\alpha) = \frac{g_{n,i}(t(\alpha))}{g_{n,i}(t(\alpha))}) \\
\approx \\
V(f_i)(\cap \mathbb{R}) \\
\rightarrow t(\alpha) \\
\leftarrow t(\alpha)
\end{align*}
\]

For computing a RUR one has to solve two problems:

- finding a separating element \( t \);
• given any polynomial $t$, compute a RUR-Candidate $f_1, g_{t,1}, g_{t, X_1}, \ldots, g_{t, X_n}$ such that
  if $t$ is a separating polynomial, then the RUR-Candidate is a RUR.

According to [26], a RUR-Candidate can explicitly be computed when knowing a suitable representation of $\mathbb{Q}[X_1, \ldots, X_n]/\mathcal{I}$:

- $f_i = \sum_{i=0}^{\mathcal{I}} a_i T^i$ is the characteristic polynomial of $m_i$. Let $\mathcal{I}$ denote by $\overline{f_i}$ its square-free part.
- for any $v \in \mathbb{Q}[X_1, \ldots, X_n]$, $g_{t,v}(T) = \sum_{i=0}^{\deg\overline{f_i}} \text{Trace}(m_{v,T}) H_{d,T-i-1}(T)$, $d = \deg(f_i)$ and $H_j(T) = \sum_{i=0}^{j} a_i T^{i+j}$.

In [26], a strategy is proposed for computing a RUR for any system (a RUR-Candidate and a separating element), but there are special cases where it can be computed differently. When $X_1$ is separating $V(\mathcal{I})$ and when $f$ is a radical ideal, the system is said to be in shape position. In such cases, the shape of the lexicographic Gröbner basis is always the following:

$$\begin{align*}
  f(X_1) &= 0 \\
  X_2 &= f_2(X_1) \\
  & \vdots \\
  X_n &= f_n(X_1)
\end{align*}$$

As shown in [26], if the system is in shape position, $g_{X_1,1} = f'_{X_1}$ and we have $f_{X_1} = f$ and $f_i(X_1) = g_{X_1, X_1}(X_1) g_{X_1, 1}(X_1) \mod f$. Thus the RUR associated with $X_1$ and the lexicographic Gröbner basis are equivalent up to the inversion of $g_{X_1, 1} = f'_{X_1}$ modulo $f_{X_1}$. In the rest of the paper we call this object a RR-Form of the corresponding lexicographic Gröbner basis. The RUR is well known to be much smaller than the lexicographic Gröbner basis in general (this may be explained by the inversion of the denominator) and thus will be our privileged symbolic output. Note that it is easy to check that a system is in shape position once knowing a RUR-Candidate (and so to check that $X_1$ separates $V(\mathcal{I})$); it is necessary and sufficient that $f_{X_1}$ is square-free.

These results have many practical drawbacks since, for most robots, the systems which model the DKP are in shape position. We thus can multiply the strategies for computing a symbolic solution: one can compute the RR-Form Gröbner directly using [8] or [11] for example or by change of ordering like in [9] or a RUR using the algorithm from [26]. Choosing the right strategy will be part of our experimental section.

### 2.5 From formal to numerical solutions

Computing a RUR reduces the resolution of a zero-dimensional system to solving one polynomial in one variable ($f_i$) and to evaluating $n$ rational fractions ($\frac{g_{X_1,1}(T)}{g_{X_1,1}(T)}, i = 1 \ldots n$) at its roots (note that if one simply wants to compute the number of real roots of the system, there is no need to consider the rational coordinates). The next task is thus to compute all
the real roots of the system (and only the real roots), providing a numerical approximation with an arbitrary precision (set by the user) of the coordinates.

The isolation of the real roots of \( f_t \) can be done using the algorithm proposed in \cite{27} : the output will be a list \( I_{f_t} \) of intervals with rational bounds such that for each real root \( \alpha \) of \( f_t \), there exists a unique interval in \( I_{f_t} \) which contains \( \alpha \). The second step consists in refining each interval in order to ensure that it does not contain any real root of \( g_{i,1} \). Since \( f_t \) and \( g_{i,1} \) are co-prime this computation is easy and we then can ensure that the rational functions can be evaluated using interval arithmetics without any cancellation of the denominator. This last evaluation is performed using multi-precision arithmetics (MPFI package - \cite{22}). As we will see in the experiments, the precision needed for the computations is poor and, moreover, the rational functions defined by the RUR are stable under numerical evaluation, even if their coefficients are huge (rational numbers), and thus this part of the computation is still efficient. For increasing the precision of the result, it is only necessary to decrease the length of the intervals in \( I_{f_t} \) which can easily be done by bisection or using a certified Newton’s algorithm. Note that is is quite simple to certify the sign of the coordinates : one simply has to compute some gcds (gcds of the numerators and of \( f_t \) and split, when necessary the RUR.

3 Algebraic modelisation

The computation of a Gröbner basis is very sensitive to the algebraic equations used to model the DKP. In this section, we show how to reduce significantly the computing time by modifying the standard algebraic equations describing the DKP.

3.1 Standard algebraic equations.

Lets introduce the nomenclature used in this part :

- \( R_f \): the base Cartesian reference frame of center \( O \).
- \( R_m \): the Cartesian reference frame of center \( C \) (end-effector) relative to the mobile platform.
- \( A_i \): location of the center of the joint of kinematics chain \( i \) on the base.
- \( B_i \): location of the center of the joint of kinematics chain \( i \) on the moving platform.
- \( L_i \): overall effective distance between \( B_i \) and \( A_i \).

Lets denote by \( \vec{v}_{i,R} \) (resp. \( M_{i,R} \)) the coordinates of the vector \( \vec{v} \) (resp. the point \( M \)) with respect to the Cartesian reference frame \( R \), and by \( N_{M,R} \) the matrix which columns are \( M_iN_{i,R}, \ i = 1 \ldots 6 \).

An hexapod shall be described by a geometric model where:
\begin{itemize}
  \item \(\overrightarrow{OA}_{i|\mathcal{R}_f}\) describe the base geometry;
  \item \(\overrightarrow{OB}_{i|\mathcal{R}_m}\) describe the mobile platform geometry;
  \item \(L_i\), \(i = 1 \ldots 6\) are the kinematics chains lengths.
\end{itemize}

According to these notations, solving the \textit{DKP} problem consists in computing \(\overrightarrow{OB}_{i|\mathcal{R}_f}\) with respect to the constraints: \(L_i^2 = ||A_iB_i||^2\), \(i = 1 \ldots 6\). Among the numerous existing \textit{DKP} formulations, there exists two main families which are commonly used in computer algebra:

\begin{itemize}
  \item \textbf{Position based equations}. Any rigid body such as the mobile platform can also be spatially located through the position of three of its distinct points such as the first three joints \(B_1, B_2\) and \(B_3\) (supposed to be distinct). Since the body is rigid, the positions of \(B_4, B_5, B_6\) and \(C\) can be deduced from the positions of \(B_1, B_2\) and \(B_3\).
  \item \textbf{Displacement based equations}. If there exists any mobile platform position \(\overrightarrow{OB}_{i|\mathcal{R}_f}\) which meets the constrains \(L_i^2 = ||A_iB_i||^2\), \(i = 1 \ldots 6\), then there exists a rotation \(\mathcal{R}\) such that:
\begin{equation}
\overrightarrow{OB}_{i|\mathcal{R}_f} = \overrightarrow{OC}_{i|\mathcal{R}_f} + \mathcal{R} \cdot \overrightarrow{CB}_{i|\mathcal{R}_m}, \quad i = 1 \ldots 6
\end{equation}
\end{itemize}

Position based equations have already been used by several authors and we describe first the ones from [15], which are already optimized for \(\text{Gröbner bases computations.}\) The systems consists in 9 equations depending on 9 variables and will be denoted by \(\mathcal{S}_{3\text{pt}}\).

As used by several authors, the natural way to set an algebraic equation system from (4) is to straightforwardly use the rotation matrix parameters and the vector \(\overrightarrow{OC}_{i|\mathcal{R}_f}\) coordinates as unknowns. Since \(\mathcal{R}\) is a rotation (a \(3 \times 3\) matrix), the following relations hold:
\begin{equation}
\mathcal{R}^T \mathcal{R} = \mathcal{I}_3, \quad \det(\mathcal{R}) = 1
\end{equation}

This first set of equations depend neither on the geometry of the platform, nor on the length of the legs and will be denoted by \(\mathcal{S}_{\text{dis},\text{rot}}\).

The constraints \(L_i^2 = ||A_iB_i||^2\), \(i = 1 \ldots 6\) can then be expressed accordingly to equations (4):
\begin{equation}
L_i^2 = ||\overrightarrow{OB}_{i|\mathcal{R}_f} - \overrightarrow{OA}_{i|\mathcal{R}_f}||^2 = ||\overrightarrow{OC}_{i|\mathcal{R}_f} + \mathcal{R} \cdot \overrightarrow{CB}_{i|\mathcal{R}_m} - \overrightarrow{OA}_{i|\mathcal{R}_f}||^2.
\end{equation}

Let denote by \(\mathcal{S}_{\text{dis},\text{geom}}\) the resulting system. The full modelisation of the displacement leads to an algebraic system \(\mathcal{S}_{\text{rot}} = \mathcal{S}_{\text{dis},\text{rot}} \cup \mathcal{S}_{\text{dis},\text{geom}}\) made of 13 polynomial equations depending on 12 unknowns \([X, Y, Z, r_{ij}, j = 1 \ldots 3, i = 1 \ldots 3]\), where \([X, Y, Z]\) are the coordinates of \(\overrightarrow{OC}_{i|\mathcal{R}_f}\). Note that a useful trick to speed up the final computation is to precompute a \(\text{Gröbner basis of }\mathcal{S}_{\text{dis},\text{rot}}\) and to add to the previous computation all the equations coming from (4).

\[\text{INRIA}\]
3.2 Quaternions

A standard way for describing the orientation of a parallel robot is to use quaternions (13, 23, 17, etc.). We propose to exploit a little bit more some blind properties of such a modeling together with specific behavior of modern algorithms for computing Gröbner bases. In short, we extract variables which are “intrinsically” linear early in the algorithm to decrease the number of variables with a significant influence on the computation times.

3.2.1 The generic case

To parameterize the group of rotations we can use the Cayley transform; if \( H \) is any anti-symmetric matrix:

\[
H = \begin{bmatrix}
0 & a & b \\
-a & 0 & c \\
-b & -c & 0
\end{bmatrix}
\]

then, provided that \( 1 \) is not an eigenvalue of \( H \), we have that

\[
\mathcal{R} = \frac{I + H}{I - H} = \begin{bmatrix}
-\frac{1+a^2+b^2}{2} & \frac{a-bc}{2} & \frac{ac+b}{2} \\
-\frac{a-bc}{2} & -\frac{1+a^2+c^2}{2} & \frac{a+b}{2} \\
\frac{ac-b}{2} & \frac{a-bc}{2} & -\frac{1+c^2+b^2}{2}
\end{bmatrix}
\]

is a rotation. Conversely, if \( \mathcal{R} \) is a rotation then \( H = \frac{\mathcal{R} - I}{\mathcal{R} + I} \) is anti-symmetric, assuming that \(-1\) is not an eigenvalue of \( \mathcal{R} \).

Expressing relation (1) and removing the denominators, one obtain a system depending on 6 variables \([X, Y, Z, a, b, c]\). We can suppose that \( \overline{CB}_1|_{n_m} = 0 \) and \( \overline{OA}_1^\top = 0 \). Then equations (8) can be rewritten:

\[
\begin{align*}
L_1^2 &= ||\overline{OC}||^2 \\
L_i^2 - L_1^2 &= ||\mathcal{R} \cdot \overline{CB}_i|_{n_m} - \overline{OA}_i^\top||^2 + 2 \left\langle \mathcal{R} \cdot \overline{CB}_i|_{n_m} - \overline{OA}_i^\top, \overline{OC} \right\rangle 
\end{align*}
\]

for \( i = 2, \ldots, 6 \).

Hence if we assume for a while that \( \mathcal{R} \) is known, then \([X, Y, Z] = \overline{OC}\) is solution of a linear system (5 equations and 3 unknowns). Therefore it is enough to compute directly the Gröbner basis of (9) for an elimination order such that \([X, Y, Z] > [a, b, c]\) and to eliminate the first block of variables (in fact the elimination is done very quickly in the first steps of the algorithm). Hence the computed Gröbner basis depends only on \( a, b, c \) and the shape of the result is usually in shape position (and thus in the form (7.4) with \( n = 3 \)). Let denote by \( S_{\text{quat}} \) the resulting system. Based on the \( F_5 \) algorithm, a specific Gröbner engine has been implemented which computes a lexicographic Gröbner basis and eliminates as soon as possible a block of \( k \) variables. As we will see in the next section the computation of such a basis is not only faster but also the size of the result (mainly the size of the coefficients) is much smaller.
3.2.2 The non generic case

Since the above formulations are valid only if $-1$ is not an eigenvalue value of $\mathcal{R}$ we need to detect and solve these exceptional situations to certify the results.

i. we first compute a Gröbner basis of $\mathcal{S}_{\text{quat}}$ for any ordering and apply theorem\footnote{A theorem in the reference material.} to count the number of complex solutions counted with multiplicities; If it is 40 or infinite we are sure that no solution is missing or that the robots is singular and we are done. Note that for singular robots it may happens that $\mathcal{S}_{\text{quat}}$ admits a finite number of solutions but that the additional linear system admits an infinite number of solutions.

ii. other else, we take the projective version of (32.1) by applying the following transform $[a \rightarrow \frac{c}{d}, b \rightarrow \frac{b}{d}, c \rightarrow \frac{c}{d}]$ and then set $d = 0$ we obtain

$$\mathcal{R}_\infty = \begin{bmatrix}
\frac{c^2-a^2-b^2}{c^2+a^2+b^2} & -2 & 0 \\
-2 & \frac{bc}{c^2+a^2+b^2} & -2 & \frac{ac}{c^2+a^2+b^2} \\
2 & \frac{ab}{c^2+a^2+b^2} & -2 & \frac{b^2+c^2-a^2}{c^2+a^2+b^2}
\end{bmatrix}$$

We thus generate an algebraic system as in the generic situation (replacing $\mathcal{R}$ by $\mathcal{R}_\infty$) and we test whether there are some solutions such that $a^2 + b^2 + c^2 \neq 0$. Note that this computation is much faster than the general computation (less than 0.1 sec for all the examples).

3.2.3 Recovering the real solutions of the DKP

If we assume that $\mathcal{S}_{\text{quat}}$ has been solved using a RR-form or a RUR and the isolation process exposed in sections\footnote{Sections references in the material.} and\footnote{Another section reference.} one needs to compute, for each real solution (rotations), the related translation $[X, Y, Z]$ which remains to solve a linear system of 5 equations in 3 variables. In a first stage, this can be tried using multi-precision interval arithmetic.

- **A** - check that at least one $3 \times 3$ minor never vanishes (by means of intervals not containing 0) at the real roots of $\mathcal{S}_{\text{quat}}$, we are done (usual situation).

- **B** - If not (exceptional situation), one symbolically computes all the $3 \times 3$ minors and, for each; check if it vanishes at some root of the system : one simply needs to add the minor to the system and solve it. Note that if such a system has a solution, the first polynomial of the RUR is a factor of the first polynomial of the RUR of $\mathcal{S}_{\text{quat}}$.

- **C** - If at least one of these minors do not vanish at any root of $\mathcal{S}_{\text{quat}}$, one increases the precision of the computed solutions and the precision of the interval arithmetic until finding the unique translations corresponding to the rotations : we are sure that such a process will end since the filter from step A will conclude that at least one of the $3 \times 3$ minor never vanishes after a finite number of tries.
• D - Else, one can ignore the roots of $S_{\text{quat}}$ which drive to an non invertible system: they correspond to singular positions of the robot, and apply the above process on the others. In practice, this can be done by removing the factor of the first polynomial of the RUR of $S_{\text{quat}}$ by the one obtained at step B and run this process again from step A.

Note that the solution can be obtained with an arbitrary precision when using multi-precision interval arithmetics such as MPFI since the solutions of $S_{\text{quat}}$ can be refined on demand accordingly to section 2.3

4 Experiments

All the computations were done on a PC Xeon (2.8 Ghz + 2 Go RAM). In the following experiments we use the following benchmark:

• Using a numerical global optimization program, Dietmaier gives explicitly an example of a robot with 40 real roots; The coordinates of the base and platform are the following:

Base:= [
[0,0,0],
[0.542805,0,0],
[0.956919,-0.528915,0],
[0.665885,-0.353482,1.402538],
[0.478359,1.158742,0.107672],
[-0.137087,-0.235121,0.353913]
];
Platform:=[
[0,0,0],
[1.107915,0,0],
[0.549094,0.756663,0],
[0.735077,-0.223935,0.525991],
[0.514188,-0.526063,-0.368418],
[0.890473,0.094733,-0.205018]
];

The 40 real solutions are obtained for the following legs lengths:

Legs:=[1,0.645275,1.086284,1.503439,1.281933,0.771071];

we show that using the tools presented in the paper it is very easy to check that the solutions are really real numbers (and not complex number with a very small
imaginary part). If \( \delta \) is any positive integer we use the following function to convert original floating point coefficients to rationals:

\[
f_\delta(x) = \frac{[2^\delta x]}{2^\delta}
\]

and thus \texttt{dietmaier.\_\_method} is the algebraic system \( \{S_{\text{quat}}, S_{\text{pt}}, S_{\text{rot}}\} \) depending on the value of \texttt{method}. We will see in that if \( \delta \) is too small then we have not 40 real roots: for instance when \( \delta = 17 \) we have only 38 real roots and two complex numbers whose imaginary parts are \( \approx \pm 0.021 \).

- **Left Hand** This example is a particular case of the "left hand" robot by Jean-Pierre Merlet. We took the following coordinates for defining its base and its platform:

```plaintext
Base:=[
[-9.704000000, 9.107000000, 0],
[9.708000000, 9.110000000, 0],
[12.7540000, 3.892000000, 0],
[599/200, -12993/1000, 3/500],
[-1503/500, -13, 1/200],
[-12.77000000, 3.901000000, 1/1000]
]

Platform:=[
[-3003/1000, 7.290000000, 0],
[601/200, 7.296000000, 0],
[7.814000000, -1.053000000, 0],
[4.812000000, -6.246000000, 3/1000],
[-4.823000000, -6.257000000, 1/125],
[-7.818000000, -1.057000000, 1/100]
]
```

Moreover, we set the length of the legs to the following values for our experiments:

```plaintext
Legs:=[9999/500,4001/200,20007/1000,
20007/1000,19993/1000,5001/250]:
```

It is interesting to test this configuration because it is a exceptional case: the system has only 36 complex solutions (and not 40). In that case, \( 1 \) is always an eigenvalue of \( R \) so that

\[
R = \begin{bmatrix}
1 & 0 & 0 \\
0 & ? & ? \\
0 & ? & ?
\end{bmatrix} = \begin{bmatrix}
1 & 0 \\
0 & R_\perp
\end{bmatrix}
\]

where \( R_\perp \) is a rotation.

Moreover we can split the 36 solutions of the problem into two sets:
i. computing with $R_\infty:-1$ is eigenvalue of $R$ (or $R_\varepsilon$) of multiplicity 2 for 6 solutions.

ii. computing with $R:-1$ is not an eigenvalue of $R$ for the other 30 complex solutions.

Left Hand $+\varepsilon$ is a perturbed version of the previous robot: it has now 40 real roots and the same number of real roots (8).

- **Spat** 66 is a generic system with 40 complex roots (the coordinates of the robot as well as the lengths of the legs are chosen randomly in $[-1000,1000]$). It has been used as a challenging example in [10] and [24]. The solution proposed in [24], based on Gröbner bases for computing the number of real roots, took several hours. By running this old strategy on the machine used for the benchmarks included in this article, we observe a computation time of 819 sec:

  - Gröbner basis for the DRL ordering using an optimized variant of burchberger’s algorithm (t-grobner) : 415 sec;
  - Univariate Representation using symmetric functions such as in [25] : 243 sec;
  - Sturm sequences for counting the real roots : 161 sec.

These results must be compared with 0.75 sec we now observe by using modern algorithm (F4/F5 for getting an RR-Form + Descartes’ method for getting the certified solutions) and the modelization of Section 472. The speed-up is about 1000.

In the following table size is twice the size of the output in the basis $b=2^{32}$; for instance hence 146 means that the biggest integer is not bigger than $2^{2^{146}} = 2^{336}$. For all but one example we give only the CPU time for computing the lexic Gröbner basis in Rational Form (denoted by RR in the table); for the first example (dietmaier_16_quat) a DRL Gröbner basis has been computed (DRL); surprisingly the computation is more difficult for a DRL Gröbner basis than for a Lexico Gröbner basis. It might seen than the size of the coefficients in the Rational Form of the Gröbner bases is smaller than in any other representation.
<table>
<thead>
<tr>
<th>Benchmark</th>
<th>r/drl</th>
<th>size</th>
<th>CPU</th>
<th>RUR</th>
<th>ISO(32)</th>
<th>ISO(64)</th>
<th>Comment</th>
</tr>
</thead>
<tbody>
<tr>
<td>dietmaier_16_quat</td>
<td>rr</td>
<td>146</td>
<td>1.0</td>
<td>-</td>
<td>0.20</td>
<td>0.28</td>
<td>40 complex roots</td>
</tr>
<tr>
<td>dietmaier_16_quat</td>
<td>dr1</td>
<td>690</td>
<td>8.7</td>
<td>6.5</td>
<td>0.46</td>
<td>0.59</td>
<td>38 real roots</td>
</tr>
<tr>
<td>dietmaier_16_3pt</td>
<td>dr1</td>
<td>842</td>
<td>13.3</td>
<td>9.4</td>
<td>0.54</td>
<td>0.75</td>
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</tr>
<tr>
<td>dietmaier_16_3pt</td>
<td>rr</td>
<td>325</td>
<td>4.0</td>
<td>-</td>
<td>0.56</td>
<td>0.71</td>
<td>&quot;</td>
</tr>
<tr>
<td>dietmaier_16_rot</td>
<td>rr</td>
<td>288</td>
<td>2.0</td>
<td>-</td>
<td>0.64</td>
<td>0.82</td>
<td>&quot;</td>
</tr>
<tr>
<td>dietmaier_18_quat</td>
<td>rr</td>
<td>165</td>
<td>1.1</td>
<td>-</td>
<td>0.21</td>
<td>0.32</td>
<td>40 real roots</td>
</tr>
<tr>
<td>dietmaier_18_3pt</td>
<td>rr</td>
<td>366</td>
<td>4.5</td>
<td>-</td>
<td>0.65</td>
<td>0.84</td>
<td>&quot;</td>
</tr>
<tr>
<td>dietmaier_18_rot</td>
<td>rr</td>
<td>326</td>
<td>2.2</td>
<td>-</td>
<td>0.75</td>
<td>0.93</td>
<td>&quot;</td>
</tr>
<tr>
<td>dietmaier_24_quat</td>
<td>rr</td>
<td>219</td>
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<td>-</td>
<td>0.25</td>
<td>0.36</td>
<td>40 real roots</td>
</tr>
<tr>
<td>dietmaier_30_quat</td>
<td>rr</td>
<td>272</td>
<td>1.8</td>
<td>-</td>
<td>0.28</td>
<td>0.40</td>
<td>&quot;</td>
</tr>
<tr>
<td>dietmaier_40_quat</td>
<td>rr</td>
<td>312</td>
<td>2.0</td>
<td>-</td>
<td>0.31</td>
<td>0.43</td>
<td>&quot;</td>
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<td>dietmaier_60_quat</td>
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<td>349</td>
<td>2.2</td>
<td>-</td>
<td>0.32</td>
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<td>dietmaier_120_quat</td>
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<td>456</td>
<td>2.9</td>
<td>-</td>
<td>0.40</td>
<td>0.56</td>
<td>&quot;</td>
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<td>Spar66_quat</td>
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<td>0.7</td>
<td>-</td>
<td>0.00</td>
<td>0.01</td>
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<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>0 real roots</td>
</tr>
<tr>
<td>LeftHand_3pt</td>
<td>rr</td>
<td>225</td>
<td>0.4</td>
<td>-</td>
<td>0.16</td>
<td>0.21</td>
<td>36 complex roots</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>8 real roots</td>
</tr>
<tr>
<td>LeftHand_rot</td>
<td>rr</td>
<td>228</td>
<td>0.6</td>
<td>-</td>
<td>0.37</td>
<td>0.39</td>
<td>&quot;</td>
</tr>
<tr>
<td>LeftHand_quat</td>
<td>rr</td>
<td>166</td>
<td>0.4</td>
<td>-</td>
<td>0.12</td>
<td>0.16</td>
<td>30 complex roots</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>8 real roots</td>
</tr>
<tr>
<td>LeftHand_rot $\mathcal{R}_\infty$</td>
<td>rr</td>
<td>14</td>
<td>0.04</td>
<td>-</td>
<td>0.00</td>
<td>0.01</td>
<td>6 complex roots</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>0 real roots</td>
</tr>
<tr>
<td>LeftHand _ quat +$\varepsilon$</td>
<td>rr</td>
<td>126</td>
<td>0.9</td>
<td>-</td>
<td>0.10</td>
<td>0.12</td>
<td>40 complex roots</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>8 real roots</td>
</tr>
</tbody>
</table>

Therefore the previous table establish a cause-and-effect relation between the size of the coefficients in the output and the CPU time for computing the corresponding object.

In [31] a floating point implantation of Gröbner bases is given; note that only an approximate Gröbner basis is given by the algorithm; the authors report a CPU time of 1067 sec (DEC Alpha 3000) to compute a lexicographic Gröbner basis. In our implementation (and a faster computer) it takes only 1 sec to compute a lexicographic Gröbner basis for the dietmaier example and ... sec to compute exactly all the real roots of the system.

The conclusion of the experiments is that the CPU time for computing a RR-Form is very stable (around 1 sec) and is minimally sensitive to configuration of the robot and the precision of the input data ($\delta$ parameter). Surprisingly, it is much more efficient to compute a lexicographical Gröbner basis or a RUR than a Gröbner basis for a DRL ordering: the simplest explanation of this behavior is that the size of the coefficients in a DRL Gröbner are bigger than in the corresponding RUR (or RR-Form).
5 Conclusion

We have improved significantly algorithms based on Gröbner bases computations for solving the DKP for parallel robots (see [24] for example) as well as the algebraic formulation of the problem. We did not compare our results with other valuable certified strategies such as homotopy methods (28, 30) or strategies based on interval analysis (see [18]) for two reasons:

- they do not solve the same kind of problems: for example, interval analysis based methods could solve robots with uncertain coordinates while we cannot; homotopy methods compute all the complex roots but cannot discriminate in a certified way the real roots from the others;
- as shown in the present paper, the modelization is of high importance, so that comparing objectively the solvers would require to find the “best” modelization for each method.

However, keeping track of the above remarks, our strategy seems to be currently the most powerfull one, with such a level of certification, for studying all the postures of an exactly knwon general robot.

The modelization proposed in section 3.2 give a precise characterization of “generic solutions” : these are those for which the linear part of the system is degenerated. A next step would be to study specifically the non generic solutions relaxing some parameters (lengths of the legs or coordinates of some joints) using recent algorithms such as [17]. A direct application would be, for example, to characterize singular robots.

Still finding all solutions is only one of the major part of the DKP. Indeed the roboticians are interested only in the current pose, i.e. the pose in which is the robot they have in front of them. Evidently this pose is in the set of the poses that is calculated but there is no known method which allows to determine which pose in the set is the current one. To show the difficulty of this problem we must define the concept of valid pose. For that purpose we consider the pose of the robot that is obtained when it is assembled for the first time (the initial assembly mode), i.e. when the 6th leg is connected to the platform, that is supposed to be known. A valid pose is defined as a pose that may be reached from the initial assembly mode by a path such that:

- the whole path must lie within the workspace of the robot (this workspace is limited by the possible change in the leg lengths and by mecahnical constraints on the joint)
- the whole path is interference-free: the legs do no intersect on the path
- the whole path is singularity-free: the jacobian of the IKP must never be singular

Hence determining is a pose is valid is equivalent to solve a robotics motion planning problem which is NP hard. There is also no known theoretical result on the problem of determining under which conditions there will be only one valid pose in the set of poses that are obtained when solving the DKP.

RR n° 5923
References


