

Robotics 1

Inverse kinematics

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Inverse kinematics what are we looking for?





direct kinematics is always unique; how about inverse kinematics for this 6R robot?

Inverse kinematics problem



- given a desired end-effector pose (position + orientation), find the values of the joint variables q that will realize it
- a synthesis problem, with input data in the form

•
$$T = \begin{bmatrix} R & p \\ 0^T & 1 \end{bmatrix} = {}^{0}A_n(q)$$
 • $r = f_r(q)$, for a task function

classical formulation:generalized formulation:inverse kinematics for a given end-effector pose T inverse kinematics for a given value r of task variables

- a typical nonlinear problem
 - existence of a solution (workspace definition)
 - uniqueness/multiplicity of solutions ($r \in \mathbb{R}^m$, $q \in \mathbb{R}^n$)
 - solution methods

Solvability and robot workspace for tasks related to a desired end-effector Cartesian pose



- primary workspace WS_1 : set of all positions p that can be reached with at least one orientation (ϕ or R)
 - out of WS₁ there is no solution to the problem
 - if $p \in WS_1$, there is a suitable ϕ (or R) for which a solution exists
- secondary (or dexterous) workspace WS₂: set of positions p that can be reached with any orientation (among those feasible for the robot direct kinematics)
 - if $p \in WS_2$, there exists a solution for any feasible ϕ (or R)

• $WS_2 \subseteq WS_1$



Workspace of Fanuc R-2000i/165F





Wrist position and E-E pose inverse solutions for an articulated 6R robot





Inverse kinematic solutions of UR10 6-dof Universal Robot UR10, with non-spherical wrist





video (slow motion)

desired pose

 $p = \begin{pmatrix} -0.2373 \\ -0.0832 \\ 1.3224 \end{pmatrix} [m]$ $R = \begin{pmatrix} \sqrt{3}/2 & 0.5 & 0 \\ -0.5 & \sqrt{3}/2 & 0 \\ 0 & 0 & 1 \end{pmatrix}$

home configuration at start $q = (0 - \pi/2 \ 0 - \pi/2 \ 0 \ 0)^{T}$ [rad]





8 inverse kinematic solutions of UR10





Multiplicity of solutions few examples



- E-E positioning of planar 2R robot (m = n = 2)
 - 2 regular solutions in $int(WS_1)$
 - 1 solution on ∂WS_1 • for $l_1 = l_2$: ∞ solutions in WS_2 } singular solutions
- E-E positioning of elbow-type spatial 3R robot (m = n = 3)
 - 4 regular solutions in WS₁ (with singular cases yet to be investigated ...)
- spatial 6R robot arms (m = n = 6)
 - \leq 16 distinct solutions, out of singularities: this "upper bound" of solutions was shown to be attained by a particular instance of "orthogonal" robot, i.e., with twist angles $\alpha_i = 0$ or $\pm \pi/2$ ($\forall i$)
 - analysis based on algebraic transformations of robot kinematics
 - transcendental equations are transformed into a single polynomial equation in one variable (number of roots = degree of the polynomial)
 - seek for a transformed polynomial equation of the least possible degree

A 6R robot with 16 IK solutions

all distinct and non-singular



Algebraic transformations whiteboard ...



start with some trigonometric equation in the joint angle θ to be solved ...

 $a \sin \theta + b \cos \theta = c$ (*)

introduce the algebraic transformation (... and the related inverse formulas)

 $u = \tan(\theta/2)$

$$\Rightarrow \quad \sin \theta = \frac{2u}{1+u^2} \quad \cos \theta = \frac{1-u^2}{1+u^2} \qquad (\Rightarrow \ \sin^2 \theta + \cos^2 \theta = 1)$$
$$\tan \theta = \tan 2(\theta/2) = \frac{2\tan(\theta/2)}{1-\tan^2(\theta/2)} = \frac{2u}{1-u^2} \qquad (\text{using the duplication formula})$$

substituting in (*)

$$a \frac{2u}{1+u^2} + b \frac{1-u^2}{1+u^2} = c \quad \Rightarrow$$
$$a + \sqrt{a^2 + b^2 - c^2}$$

$$\Rightarrow \quad u_{1,2} = \frac{u \pm vu + b - c}{b + c}$$

polynomial equation of second degree in u $(b + c) u^2 - 2a u - (b - c) = 0$

$$\Rightarrow \quad \theta_{1,2} = 2 \arctan(u_{1,2})$$

only if argument is real, else no solution

A planar 3R arm workspace and number/type of inverse solutions







Multiplicity of solutions summary of the general cases



- if m = n
 - ∄ solutions
 - a finite number of solutions (regular/generic case)
 - "degenerate" solutions: infinite or finite set, but anyway different in number from the generic case (singularity)
- if m < n (robot is kinematically redundant for the task)
 - ∄ solutions
 - ∞^{n-m} solutions (regular/generic case)
 - a finite or infinite number of singular solutions
- use of the term singularity will become clearer when dealing with differential kinematics
 - instantaneous velocity mapping from joint to task velocity
 - lack of full rank of the associated $m \times n$ Jacobian matrix J(q)

Dexter 8R robot arm



- m = 6 (position and orientation of E-E)
- n = 8 (all revolute joints)
- ∞^2 inverse kinematic solutions (redundancy degree = n m = 2)



exploring inverse kinematic solutions by a robot self-motion

Solution methods



ANALYTICAL solution (in closed form)



NUMERICAL solution (in iterative form)

- preferred, if it can be found*
- use ad-hoc geometric inspection
- algebraic methods (solution of polynomial equations)
- systematic ways for generating a reduced set of equations to be solved

* sufficient conditions for 6-dof arms

- 3 consecutive rotational joint axes are incident (e.g., spherical wrist), or
- 3 consecutive rotational joint axes are parallel
- D. Pieper, PhD thesis, Stanford University, 1968

certainly needed if n > m
 (redundant case) or at/close to singularities

- slower, but easier to be set up
- in its basic form, it uses the (analytical) Jacobian matrix of the direct kinematics map

$$J_r(q) = \frac{\partial f_r(q)}{\partial q}$$

 Newton method, Gradient method, and so on...

Inverse kinematics of planar 2R arm





direct kinematics $p_x = l_1c_1 + l_2c_{12}$ $p_y = l_1s_1 + l_2s_{12}$

data q_1, q_2 unknowns

"squaring and summing" the equations of the direct kinematics

$$p_x^2 + p_y^2 - (l_1^2 + l_2^2) = 2l_1l_2(c_1c_{12} + s_1s_{12}) = 2l_1l_2c_2$$

and from this

is outside robot workspace!)









Algebraic solution for q_1

 $p_x = l_1c_1 + l_2c_{12} = l_1c_1 + l_2(c_1c_2 - s_1s_2)$ $p_y = l_1s_1 + l_2s_{12} = l_1s_1 + l_2(s_1c_2 + c_1s_2)$ another linear in solution s_1 and c_1 method... $\begin{bmatrix} l_1 + l_2 c_2 & -l_2 s_2 \\ l_2 s_2 & l_1 + l_2 c_2 \end{bmatrix} \begin{bmatrix} c_1 \\ s_1 \end{bmatrix} = \begin{bmatrix} p_x \\ p_y \end{bmatrix}$ except if $l_1 = l_2$ and $c_2 = -1$ $\det = l_1^2 + l_2^2 + 2l_1l_2c_2 > 0$ being then q_1 undefined (singular case: ∞^1 solutions) $q_1 = \operatorname{atan2}\{s_1, c_1\}$

 $= \operatorname{atan2}\{(p_y(l_1 + l_2c_2) - p_xl_2s_2)/\operatorname{det}, (p_x(l_1 + l_2c_2) + p_yl_2s_2)/\operatorname{det}\}$

notes: a) this method provides directly the result in $(-\pi, \pi]$

b) when evaluating atan2, det > 0 can be in fact eliminated from the expressions of s_1 and c_1 (not changing the result)



Inverse kinematics of polar (RRP) arm





Inverse kinematics of 3R elbow-type arm





symmetric structure without offsets e.g., first 3 joints of Mitsubishi PA10 robot

 $WS_1 = \{ \text{spherical shell centered at } (0,0,d_1), \\ \text{with outer radius } R_{out} = L_2 + L_3 \\ \text{and inner radius } R_{in} = |L_2 - L_3| \}$

4 regular inverse kinematics solutions in WS_1

more details (e.g., full handling of singular cases) can be found in the solution of Exercise #1 in written exam of 11 Apr 2017

Inverse kinematics of 3R elbow-type arm step 1



Inverse kinematics of 3R elbow-type arm step 2





Inverse kinematics of 3R elbow-type arm step 3





combine first the two equations of direct kinematics and rearrange the last one

$$\begin{aligned} c_1 p_x + s_1 p_y &= L_2 c_2 + L_3 c_{23} \\ &= (L_2 + L_3 c_3) c_2 - L_3 s_3 s_2 \\ p_z - d_1 &= L_2 s_2 + L_3 s_{23} \\ &= L_3 s_3 c_2 + (L_2 + L_3 c_3) s_2 \end{aligned}$$

define and solve a linear system Ax = bin the algebraic unknowns $x = (c_2, s_2)$

4 regular solutions for q₂,
depending on the combinations of {+, -} from q₁ and q₃

$$q_{2}^{\{\{f,b\},\{u,d\}\}}$$

$$= \operatorname{atan2}\left\{s_{2}^{\{\{f,b\},\{u,d\}\}}, c_{2}^{\{\{f,b\},\{u,d\}\}}\right\}$$

Inverse kinematics for robots with spherical wrist





6R robot Unimation PUMA 600



8 different (regular) inverse solutions that can be found in closed form

Joint

 $a_2 = 17.000$

 $d_1 = 4.937$

spherical

wrist

25.26

a٩

~ 90°

90°

90°

~ 90°

 $a_1 = 0.75$

 $d_{4} = 17.000$

so that
$$O_6 = W$$
 directly

here d = 0

Finding nice kinematic relations whiteboard ...



the most complex inverse kinematics that can be solved in principle in closed form (i.e., analytically) is that of a 6R serial manipulator, with arbitrary DH table
 ways to systematically generate equations from the direct kinematics that could be easier to solve ⇒ some scalar equations may contain perhaps a single unknown variable!

method used for the Unimation PUMA 600 in (*) ${}^{0}T_{6} = {}^{0}A_{1}(\theta_{1}) {}^{1}A_{2}(\theta_{2}) \cdots {}^{5}A_{6}(\theta_{6}) = U_{0}$

$${}^{0}A_{1}^{-1} {}^{0}T_{6} = U_{1} (= {}^{1}A_{2} \cdots {}^{5}A_{6})$$

$${}^{1}A_{2}^{-1} {}^{0}A_{1}^{-1} {}^{0}T_{6} = U_{2} (= {}^{2}A_{3} \cdots {}^{5}A_{6})$$
or also ...
$${}^{0}T_{6} {}^{5}A_{6}^{-1} = V_{5} (= {}^{0}A_{1} \cdots {}^{4}A_{5})$$

$${}^{0}T_{6} {}^{5}A_{6}^{-1} {}^{4}A_{5}^{-1} = V_{4} (= {}^{0}A_{1} \cdots {}^{3}A_{4})$$
...
$${}^{4}A_{5}^{-1} \cdots {}^{1}A_{2}^{-1} {}^{0}A_{1}^{-1} {}^{0}T_{6} = U_{5} (= {}^{5}A_{6})$$

$${}^{0}T_{6} {}^{5}A_{6}^{-1} {}^{4}A_{5}^{-1} \cdots {}^{1}A_{2}^{-1} = V_{1} (= {}^{0}A_{1})$$

(*) Paul, Shimano, and Mayer: IEEE Transactions on Systems, Man, and Cybernetics, 1981

• generating from the direct kinematics a reduced set of equations to be solved (setting w.l.o.g. $d_1 = d_6 = 0$) \Rightarrow 4 compact scalar equations in the 4 unknowns $\theta_2, ..., \theta_5$

Manseur and Doty: International Journal of Robotics Research, 1988

Numerical solution of inverse kinematics problems



- use when a closed-form solution q to $r_d = f_r(q)$ does not exist or is "too hard" to be found
- all methods are iterative and need the matrix $J_r(q) = \frac{\partial f_r(q)}{\partial q}$ (analytical Jacobian)
- Newton method (here only for m = n, at the *k*th iteration)

•
$$r_d = f_r(q) = f_r(q^k) + J_r(q^k)(q - q^k) + o(||q - q^k||)$$

 \leftarrow neglected in Taylor expansion

$$q^{k+1} = q^k + J_r^{-1}(q^k) \left[r_d - f_r(q^k) \right]$$

- convergence for q^0 (initial guess) close enough to some q^* : $f_r(q^*) = r_d$
- problems near singularities of the Jacobian matrix $J_r(q)$
- in case of robot redundancy (m < n), use the pseudoinverse $J_r^{\#}(q)$
- has quadratic convergence rate when near to a solution (fast!)

Operation of Newton method



- in the scalar case, also known as "method of the tangent"
- for a differentiable function f(x), find a root x* of f(x*) = 0 by iterating as



animation from http://en.wikipedia.org/wiki/File:NewtonIteration_Ani.gif

Numerical solution of inverse kinematics problems (cont'd)



- Gradient method (max descent)
 - minimize the error function

$$H(q) = \frac{1}{2} \|r_d - f_r(q)\|^2 = \frac{1}{2} (r_d - f_r(q))^T (r_d - f_r(q))$$
$$q^{k+1} = q^k - \alpha \, \nabla_q H(q^k)$$

from

$$\nabla_q H(q) = \left(\frac{\partial H(q)}{\partial q}\right)^T = -\left(\left(r_d - f_r(q)\right)^T \left(\frac{\partial f_r(q)}{\partial q}\right)\right)^T = -J_r^T(q)(r_d - f_r(q))$$

we get

$$q^{k+1} = q^k + \alpha J_r^T(q^k) \big(r_d - f_r(q^k) \big)$$

 the scalar step size α > 0 should be chosen so as to guarantee a decrease of the error function at each iteration: too large values for α may lead the method to "miss" the minimum

when the step size is too small, convergence is extremely slow Robotics 1



Revisited as a feedback scheme



 $e = r_d - f_r(q) \rightarrow 0 \iff \text{closed-loop equilibrium } e = 0$ is asymptotically stable

 $V = \frac{1}{2}e^T e \ge 0$ is a Lyapunov candidate function $\dot{V} = e^T \dot{e} = e^T \frac{d}{dt} (r_d - f_r(q)) = -e^T J_r(q) \dot{q} = -e^T J_r(q) J_r^T(q) e \le 0$

Properties of Gradient method



- computationally simpler: use the Jacobian transpose, rather than its (pseudo)inverse
- same use also for robots that are redundant (n > m) for the task
- may not converge to a solution, but it never diverges
- the discrete-time evolution of the continuous scheme

$$q^{k+1} = q^k + \Delta T J_r^T(q^k) (r_d - f_r(q^k)), \quad \alpha = \Delta T$$

is equivalent to an iteration of the Gradient method

• the scheme can be accelerated by using a gain matrix K > 0

$$\dot{q} = J_r^T(q) Ke = J_r^T(q) K(r_d - f_r(q))$$

note: $K \to K + K_s$, with K_s skew-symmetric, can be used also to "escape" from being stuck in a **stationary point** of $V = \frac{1}{2}e^T K e$, by **rotating** the error Ke out of the null space of J_r^T (when a **singularity** is encountered)

A case study

analytic expressions of Newton and gradient iterations



- 2R robot with $l_1 = l_2 = 1$, desired end-effector position $r_d = p_d = (1,1)$
- direct kinematic function and error

$$f_r(q) = \begin{pmatrix} c_1 + c_{12} \\ s_1 + s_{12} \end{pmatrix} \qquad e = p_d - f_r(q) = \begin{pmatrix} 1 \\ 1 \end{pmatrix} - f_r(q)$$

Jacobian matrix

$$J_r(q) = \frac{\partial f_r(q)}{\partial q} = \begin{pmatrix} -(s_1 + s_{12}) & -s_{12} \\ c_1 + c_{12} & c_{12} \end{pmatrix}$$

Newton versus Gradient iteration

$$det J_{r}(q) = q^{k} + \begin{cases} \frac{1}{s_{2}} \begin{pmatrix} c_{12} & s_{12} \\ -(c_{1}+c_{12}) & -(s_{1}+s_{12}) \end{pmatrix}_{|q=q^{k}} \\ \alpha \begin{pmatrix} -(s_{1}+s_{12}) & c_{1}+c_{12} \\ -s_{12} & c_{12} \end{pmatrix}_{|q=q^{k}} \\ \chi \begin{pmatrix} 1 - (c_{1}+c_{12}) \\ 1 - (s_{1}+s_{12}) \end{pmatrix}_{|q=q^{k}} \\ \chi \begin{pmatrix} 1 - (s_{1}+s_{12}) \\ 1 - (s_{1}+s_{12}) \end{pmatrix}_{|q=q^{k}} \end{cases}$$

Error function



• 2R robot with $l_1 = l_2 = 1$ and desired end-effector position $p_d = (1,1)$



Configuration space of 2R robot whiteboard ...



■ can we represent the correct "distance" between two configurations q' and q'' of this robot on a (square) region in ℝ²?



• configuration space is a torus $SO(1) \times SO(1)$, i.e., the surface of a "donut"



• the right metric is a geodesic on the torus ...

Error reduction by Gradient method



- flow of iterations along the negative (or anti-) gradient
 two possible appears on stuck (at zero gradient)
- two possible cases: convergence or stuck (at zero gradient)



Convergence analysis

when does the gradient method get stuck?

- lack of convergence occurs when
 - the Jacobian matrix $J_r(q)$ is not full rank (the robot is in a "singular configuration")
 - AND the error *e* is in the null space of $J_r^T(q)$



Issues in implementation



- initial guess q^0
 - only one inverse solution is generated for each guess
 - multiple initializations for obtaining other solutions
- optimal step size $\alpha > 0$ in Gradient method
 - a constant step may work good initially, but not close to the solution (or vice versa)
 - an adaptive one-dimensional line search (e.g., Armijo's rule) could be used to choose the best α at each iteration
- stopping criteria

Cartesian error (possibly, separate for $||r_d - f_r(q^k)|| \le \varepsilon$ algorithm increment $||q^{k+1} - q^k|| \le \varepsilon_q$ position and orientation)

understanding closeness to singularities

good numerical conditioning $\sigma_{\min}\{J_r(q^k)\} \ge \sigma_0$ of Jacobian matrix (SVD) (or a simpler test on its determinant, for m = n)

Numerical tests on RRP robot

- RRP/polar robot: desired E-E position $r_d = p_d = (1, 1, 1)$ —see slide #22, with $d_1 = 0.5$
- the two (known) analytical solutions, with $q_3 \ge 0$, are

 $q^* = (0.7854, 0.3398, 1.5)$

$$q^{**} = (q_1^* - \pi, \pi - q_2^*, q_3^*) = (-2.3562, 2.8018, 1.5)$$

- norms $\varepsilon = 10^{-5}$ (max Cartesian error), $\varepsilon_q = 10^{-6}$ (min joint increment)
- $k_{max} = 15 \pmod{\#}$ iterations), $|\det J_r(q)| \le 10^{-4} \pmod{\#}$
- numerical performance of Gradient (with different steps α) vs. Newton
 - test 1: $q^0 = (0, 0, 1)$ as initial guess
 - test 2: $q^0 = (-\pi/4, \pi/2, 1)$ "singular" start, since $c_2 = 0$ (see slide #22)
 - test 3: $q^0 = (0, \pi/2, 0)$ "doubly singular" start, since also $q_3 = 0$
 - solution and plots with MATLAB code





• test 1: $q^0 = (0, 0, 1)$ as initial guess; evolution of the error norm





• test 1: $q^0 = (0, 0, 1)$ as initial guess; evolution of joint variables





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Final remarks



- an efficient iterative scheme can be devised by combining
 - initial iterations using Gradient ("sure but slow", linear convergence rate)
 - switch then to Newton method (quadratic terminal convergence rate)
- joint range limits are considered only at the end
 - check if the solution found is feasible, as for analytical methods
- or, an optimization criterion and/or constraints included in the search
 - drive iterations toward an inverse kinematic solution with nicer properties
- if the problem has to be solved on-line
 - execute iterations and associate an actual robot motion: repeat steps at times t_0 , $t_1 = t_0 + T$, ..., $t_k = t_{k-1} + T$ (e.g., every T = 40 ms)
 - a "good" choice for the initial guess q^0 at t_k is the solution of the previous problem at t_{k-1} (provides continuity, requires only 1-2 Newton iterations)
 - crossing of singularities and handling of joint range limits need special care
- Jacobian-based inversion schemes are used also for kinematic control, moving along/tracking a continuous task trajectory $r_d(t)$