

Robotics 2

Kinematic calibration

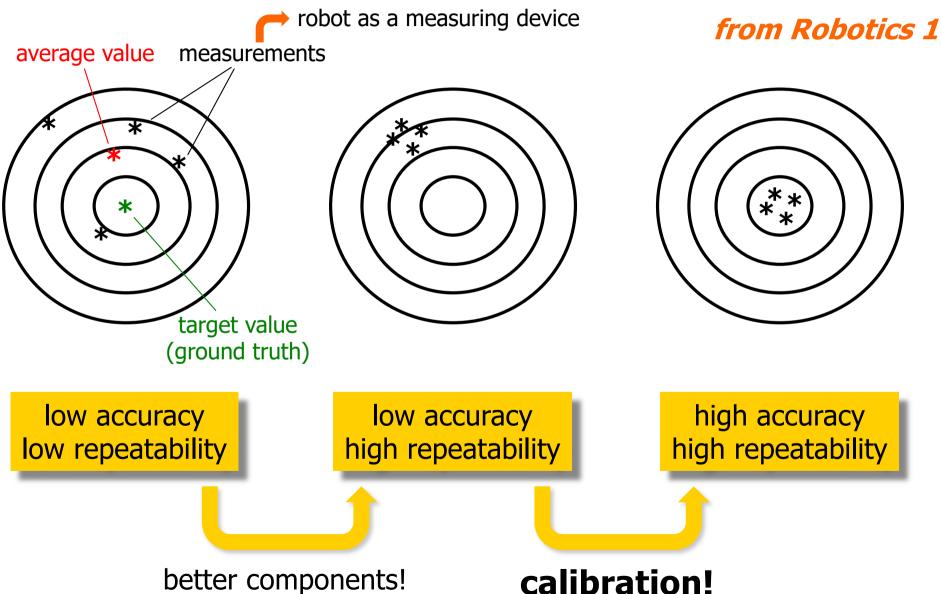
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DIPARTIMENTO DI INGEGNERIA INFORMATICA AUTOMATICA E GESTIONALE ANTONIO RUBERTI



Accuracy and Repeatability





Direct kinematics



nominal set of Denavit-Hartenberg (D-H) parameters

$$\pmb{\alpha} = \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix} \; \pmb{a} = \begin{pmatrix} a_1 \\ \vdots \\ a_n \end{pmatrix} \; \pmb{d} = \begin{pmatrix} d_1 \\ \vdots \\ d_n \end{pmatrix} \; \text{for simplicity, suppose an all-revolute joints manipulator}$$

nominal direct kinematics

$$r_{nom} = f(\alpha, \alpha, d, \theta)$$

 θ_i 's typically measured by encoders

$$r_{nom} = f(\alpha, a, d, \theta)$$
 $\theta = \begin{pmatrix} \theta_1 \\ \vdots \\ \theta_n \end{pmatrix}$

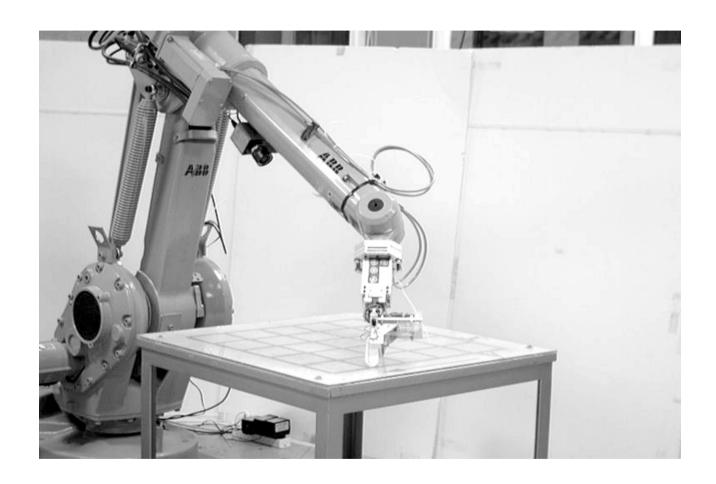
Need for calibration



- tolerances in mechanical construction and in assembly of links/joints imply small errors in actual end-effector pose (real ≠ nominal parameters)
- encoder mounting on motor axes may not be consistent with the "zero reference" of the robot direct kinematics (joint angle measures are constantly biased)
- errors distributed "along" the arm are amplified, due to the open kinematic chain structure of most robots
- calibration goal: recover as much as possible E-E pose errors by correcting the nominal set of D-H parameters, based on independent external (accurate!) measurements
- experiments to be done once for each robot, before starting operation... (and maybe repeated from time to time)



Cartesian measurement systems - 1

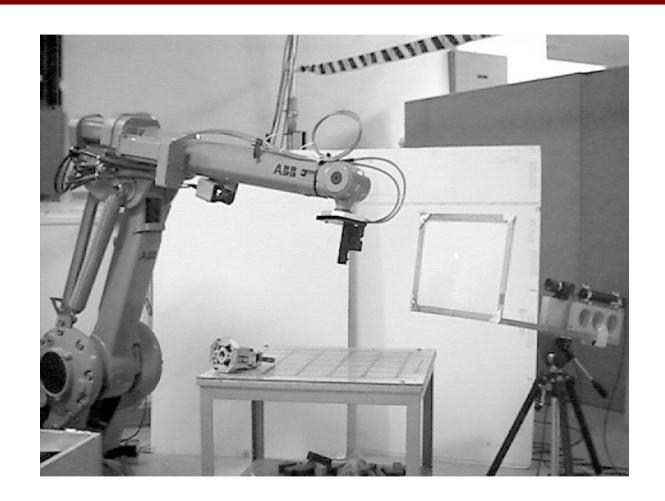


calibration table

Robotics 2 5



Cartesian measurement systems - 2



laser/camera system + triangulation

Robotics 2 6

Cartesian measurement systems - 3



FANUC 6R robot M-710iC/50

M₁ M₂

API laser tracker III www.apisensor.com

3 SMRs (Spherically-Mounted Reflectors)

laser tracker + targets on end-effector

Acquiring data for calibration



ABB

robot

4 SMRs

IRB 1600

FARO ION laser tracker



video @CoRo Lab ETS Montréal



Linearization of direct kinematics



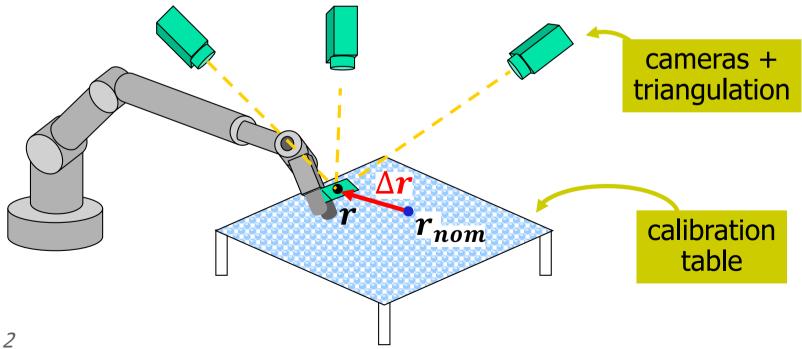
partial Jacobians evaluated in nominal conditions

$$\Delta r = r - r_{nom} = \frac{\partial f}{\partial \alpha} \cdot \Delta \alpha + \frac{\partial f}{\partial a} \cdot \Delta \alpha + \frac{\partial f}{\partial d} \cdot \Delta d + \frac{\partial f}{\partial \theta} \cdot \Delta \theta$$

"small" errors

obtained by external measurement system

first-order variations





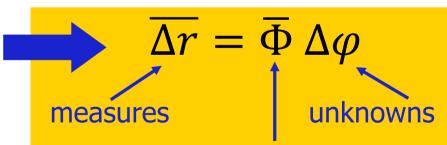
Calibration equation

$$\Delta \varphi = \begin{pmatrix} \Delta \alpha \\ \Delta a \\ \Delta d \\ \Delta \theta \end{pmatrix} \quad \Phi = \begin{pmatrix} \frac{\partial f}{\partial \alpha} & \frac{\partial f}{\partial a} & \frac{\partial f}{\partial d} & \frac{\partial f}{\partial \theta} \end{pmatrix} \qquad \qquad \Delta r = \Phi \Delta \varphi$$

$$\Delta r = \Phi$$

$$\frac{\delta\ell \times 1}{\Delta r} = \begin{pmatrix} \Delta r_1 \\ \Delta r_2 \\ \vdots \\ \Delta r_{\ell} \end{pmatrix} \qquad \frac{\delta\ell \times 4n}{\overline{\Phi}} = \begin{pmatrix} \Phi_1 \\ \Phi_2 \\ \vdots \\ \Phi_{\ell} \end{pmatrix}$$

 ℓ experiments ($\ell \gg n$)

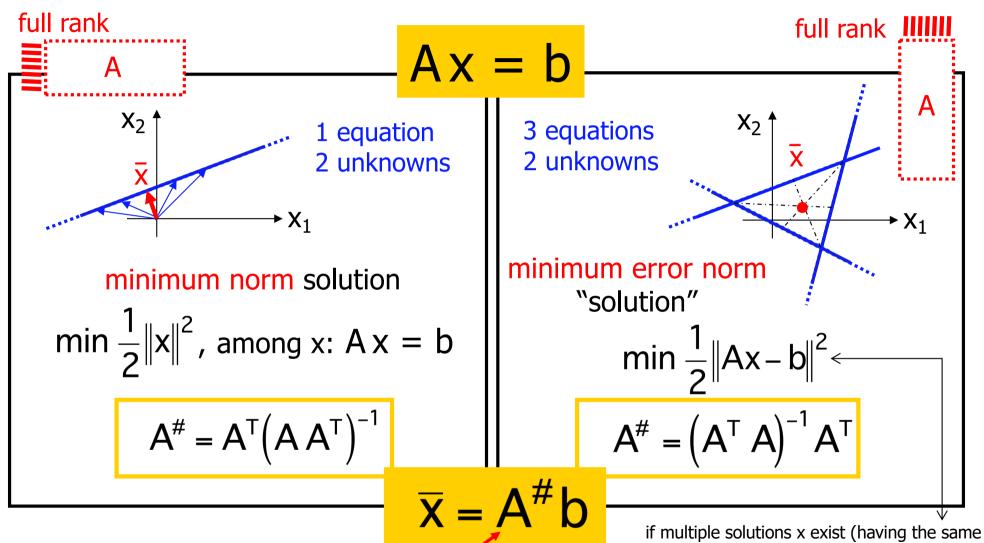


regressor matrix evaluated at nominal parameters

full column rank (for sufficiently large ℓ)

Under- and over-determined systems of linear equations





pseudoinversé!!

minimum error norm), the pseudoinverse

provides the one having **minimum norm**



Calibration algorithm

$$\begin{array}{l} \epsilon_r = 10^{-4}, \ i_{max} = 50 \\ \varphi^{(0)} = \varphi_{nom} \\ \overline{\Phi}^{(0)} = \overline{\Phi}(\varphi^{(0)}) \\ \overline{\Delta r}^{(0)} = \overline{\Delta r}(\varphi^{(0)}) \\ i = 0 \\ \text{linearized least squares problem} \\ i = 0 \\ \text{needs an iterative solution} \\ if \ \left\| \overline{\Delta r}^{(i)} \right\| \leq \epsilon_r \\ \varphi^* = \varphi^{(i)} \\ \text{final solution} \\ else \ if \ i \leq i_{max} \\ \Delta \varphi^{(i)} = (\overline{\Phi}^{(i)})^\# \, \overline{\Delta r}^{(i)} = \left(\overline{\Phi}^{(i)}^T \overline{\Phi}^{(i)}\right)^{-1} \overline{\Phi}^{(i)}^T \, \overline{\Delta r}^{(i)} \\ \varphi^{(i+1)} = \varphi^{(i)} + \Delta \varphi^{(i)} \\ \overline{\Phi}^{(i+1)} = \overline{\Phi}(\varphi^{(i+1)}) \\ \overline{\Delta r}^{(i+1)} = \overline{\Delta r}(\varphi^{(i+1)}) \\ i = i+1 \\ else \\ disp('no \ convergence \ in \ i_{max} \ iterations') \\ end \ if \end{array}$$

Improvement by kinematic calibration



- ABB IRB 120 6R industrial robot
- 1000 random configurations (collision-free by simulation)
- 50 arbitrary configurations used for measurement in calibration
- 950 configurations used for validation
- Cartesian position errors

	before calibration	after calibration
Average	1.746 mm	0.193 mm
Median	1.567 mm	0.180 mm
Standard Deviation	n 1.043 mm	0.085 mm
Min	0.050 mm	0.010 mm
Max	4.423 mm	0.516 mm

Improvement by a factor 8÷10

Final comments



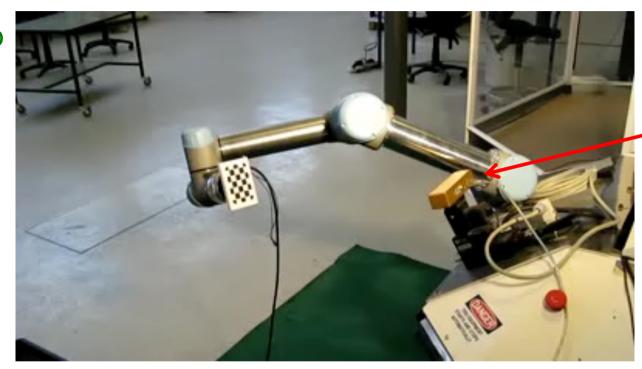
- an iterative least squares method
 - original problem is nonlinear in the unknowns, then linearized using firstorder Taylor expansion
- it is useful to calibrate first and separately those quantities that are less accurate (typically, the encoder bias)
 - keeping the remaining ones at their nominal values
- alternative kinematic descriptions can be used
 - more complex than D-H parameters, but leading to a better numerical conditioning of the regressor matrix in calibration algorithm
 - one such description uses the POE (Product Of Exponential) formula
- more in general, 6 base parameters should also be included
 - to locate 0-th robot frame w.r.t. world coordinate frame (of external sensor)
- accurate calibration/estimation of real parameters is a general problem in robotics (and beyond...)
 - for sensors (e.g., camera calibration)
 - for models (identification of dynamic parameters of a manipulator)

Calibration experiment

in a research environment



video



Videre Design stereovision camera

- automatic data acquisition for simultaneous calibration of
 - robot-camera transformation
 - DH parameters of the manipulator

Calibration experiment

in an industrial setting



FANUC
3D Laser
calibration
(with iR Vision)



video

