

#1 Inverse geometry

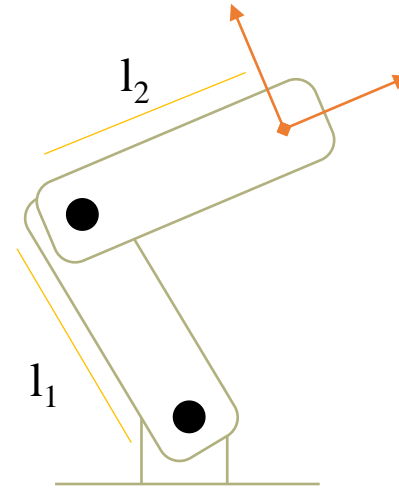
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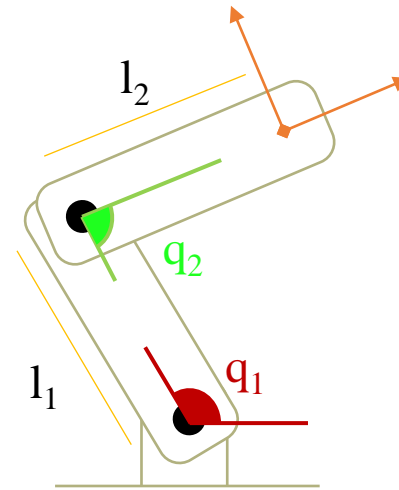
Geometry model

Robot configuration q



Geometry model

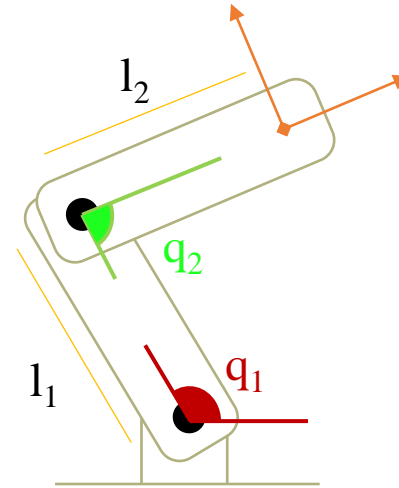
Robot configuration q



Geometry model

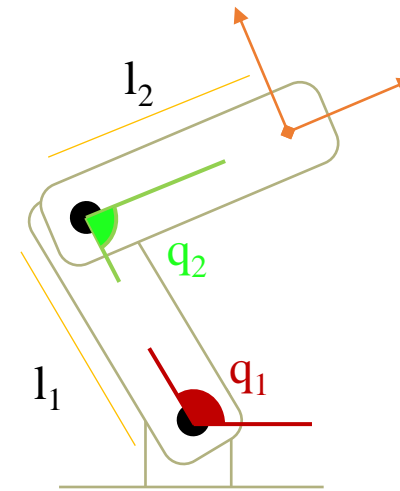
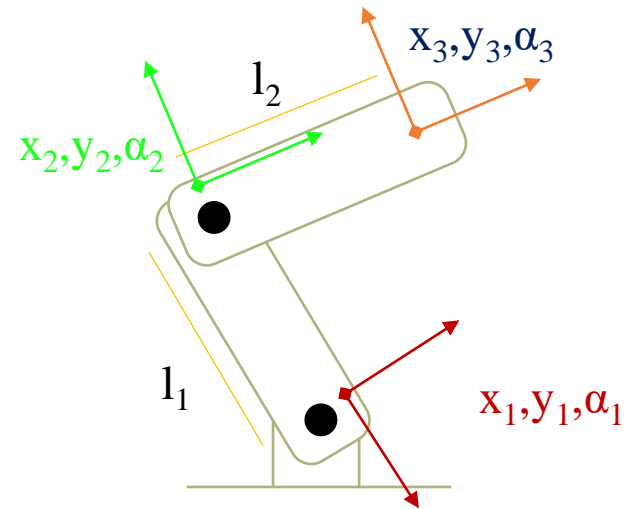
Robot configuration q

$$\begin{bmatrix} l_1 \cos(q_1) + l_2 \cos(q_1+q_2) \\ l_1 \sin(q_1) + l_2 \sin(q_1+q_2) \end{bmatrix}$$



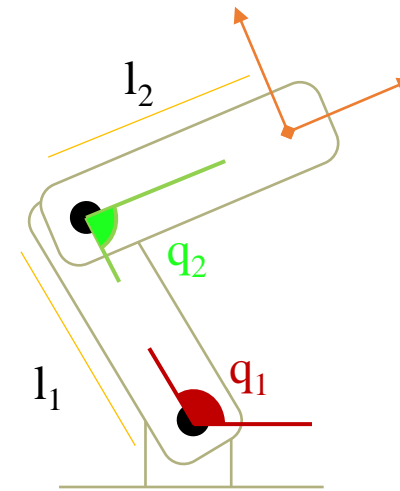
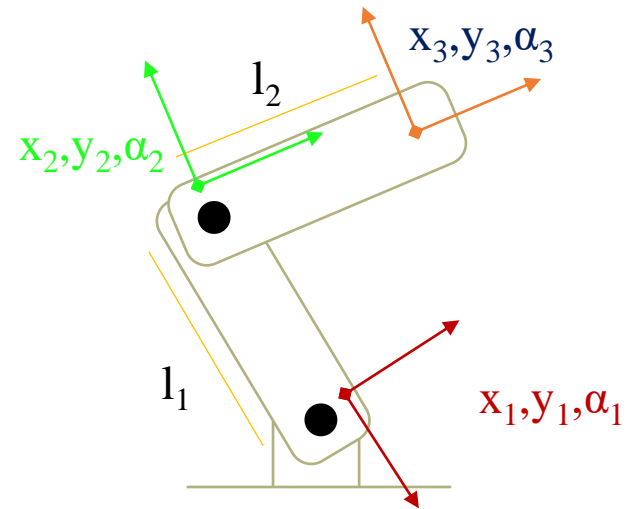
Geometry model

Robot configuration q



Geometry model

Robot configuration q



... such that

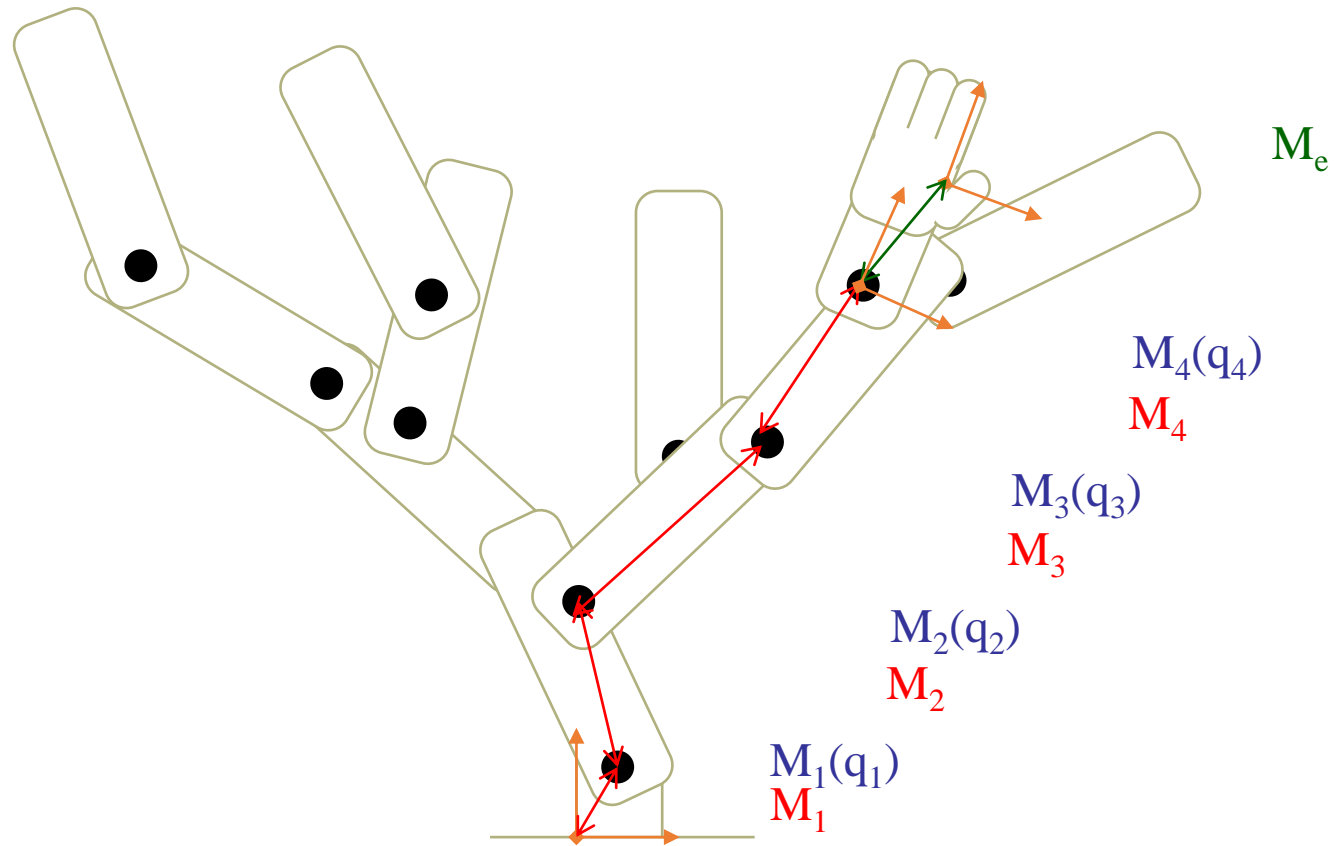
$$(x_1 - x_2)^2 + (y_1 - y_2)^2 = \text{cst}$$

$$(x_2 - x_3)^2 + (y_2 - y_3)^2 = \text{cst}$$

... etc ...

Direct geometry

The geometric model is a tree of joints and bodies



$$M(q) = M_1 \oplus M_1(q_1) \oplus M_2 \oplus \dots \oplus M_4 \oplus M_4(q_4) \oplus M_e$$

About representation of motion

The geometric model is a tree of joints and bodies

$$M = \begin{bmatrix} R & p \\ 0 & 1 \end{bmatrix}$$

Homogeneous matrix
... represents SE(3)

$$\dot{R} = \omega \times R$$

Canonical definition
of angular velocity

What is $M \in \text{SE}(3)$?

What is \dot{M} (and \dot{R})?

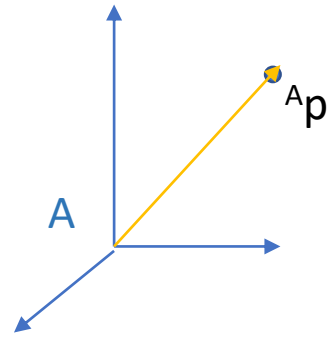
Links with the differential geometry?

$$M(q) = M_1 \oplus M_1(q_1) \oplus M_2 \oplus \dots \oplus M_4 \oplus M_4(q_4) \oplus M_e$$

Representation!



This is a point



This is not a point
This is the representation of a point

Rotation

- Rotation matrices

$$R = \begin{pmatrix} r_{00} & r_{01} & r_{02} \\ r_{10} & r_{11} & r_{12} \\ r_{20} & r_{21} & r_{22} \end{pmatrix}$$

- Derivation of a matrix

$$\dot{R} = \dots$$

Angular velocity / Angle vector

- Formal definition

$$\dot{R} = \omega \times R$$

- From rotation to velocity

- $R \rightarrow \omega$

$$R = \exp(\omega \times)$$

- From velocity to rotation?

- $\omega \rightarrow R$... integrate

$$\omega \times = \log(R)$$

- Meaning of ω : angular velocity
- Angle axis representation ($\omega = u\theta$, with u the axis, θ the angle)
- Quaternions... see Joan Sola ☺

Quaternions

- Start from complex

$1, i, -1, -i, 1 \dots$

$X, Y, -X, -Y, X \dots$

Complexes can map the 2D plan , and the 2D rotation

- Hamilton (again!) says: let's do it more complex

j so that $j^2 = -1$ and $ij = -ji$

$ij = k, jk = i \dots$

$x \mathbf{G} y = z, y \mathbf{G} z = x \dots$

- Unit quaternions map 3D rotations

$q = [w, x, y, z] = \cos(\alpha/2), \sin(\alpha/2)[a,b,c]$

Joint models

- Maps
 - From configuration space
 - To SE3 space

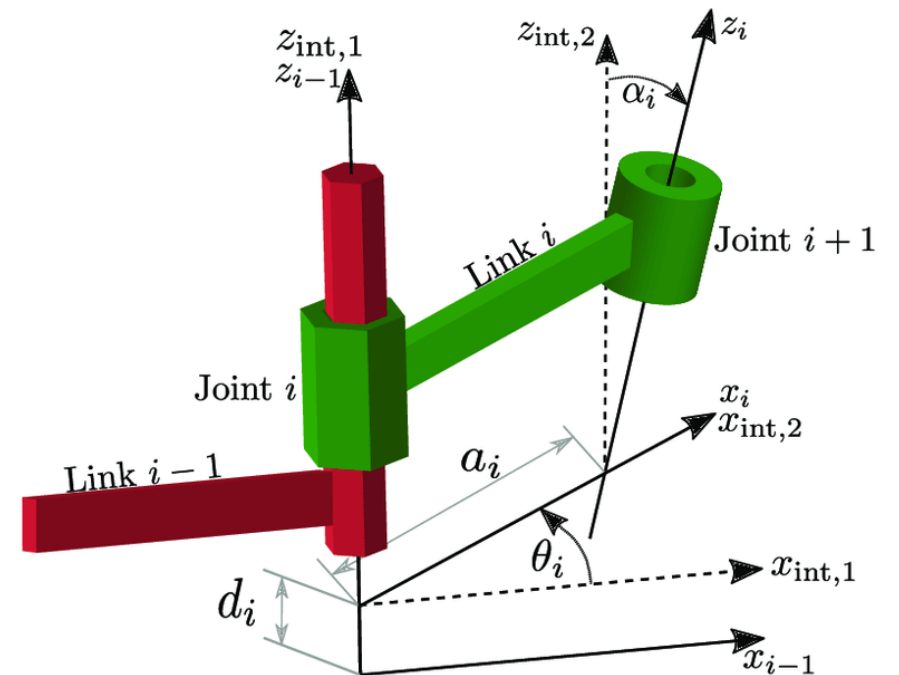
$$h(q) = {}^kM_{k+1}(q) \in SE(3)$$

- For example, Revolute-Z is:

$$h(q) = \begin{bmatrix} \cos q & \sin q & 0 & 0 \\ -\sin q & \cos q & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Kinematic model parametrization

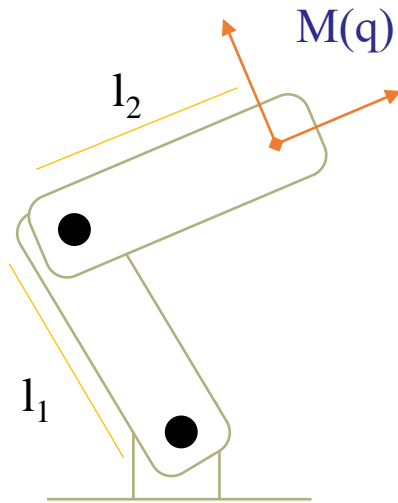
- Parent-to-child joint transformation
- Modern solution:
 - URDF model
 - with your favorite SE3 representation
- Good-old days:
with Denavit-Hartenberg
minimal parameters



Recap

| | |
|---|---|
| <code>pin.Model</code> | Kinematic tree, frame placement |
| <code>pin.Data</code> | Buffers for algorithms computations |
| <code>data.oMi</code> | Placement of the joints wrt world |
| <code>data.oMf</code> | Placement of frames wrt world |
| <code>pin.framesForwardKinematics(model, data, q, v)</code> | |
| <code>M = pin.SE3(R, p)</code> | Placement (rotation+translation) matrix |
| <code>M.rotation, M.translation</code> | |
| <code>pin.log6(M).vector</code> | SE3 log, convert placement to motion |
| <code>pin.exp3(np.array([1, 2, 3]))</code> | S03 exponential, "integrate" velocity to rotation matrix |

Inverse geometry



Being given a M^* ...

what is q such that $M(q) = M^*$

$$M^{-1}: M^* \rightarrow q = M^{-1}(M^*)$$

Numerical inversion of the geometry

- Computing analytically h^{-1} is difficult and tedious
- We can compute it numerically!

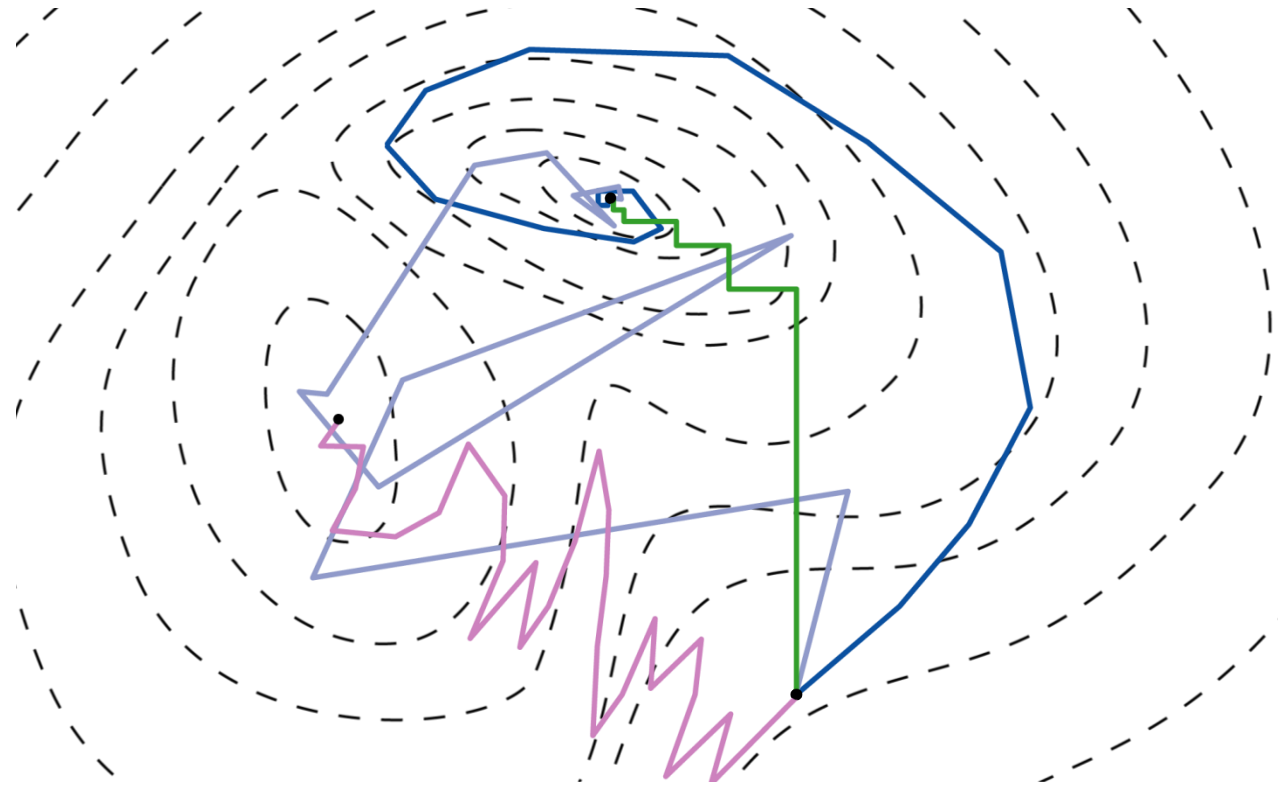
- Problem definition

$$\textit{search } f(x) - f^* = 0$$

$$\textit{min } \| f(x) - f^* \|^2$$

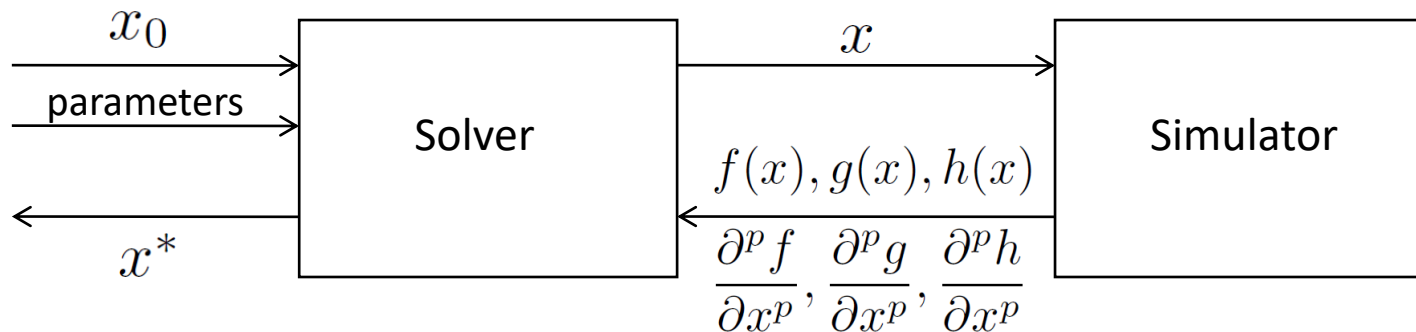
Follow the slope

- Decreasing sequence: $f(x_{k+1}) < f(x_k)$



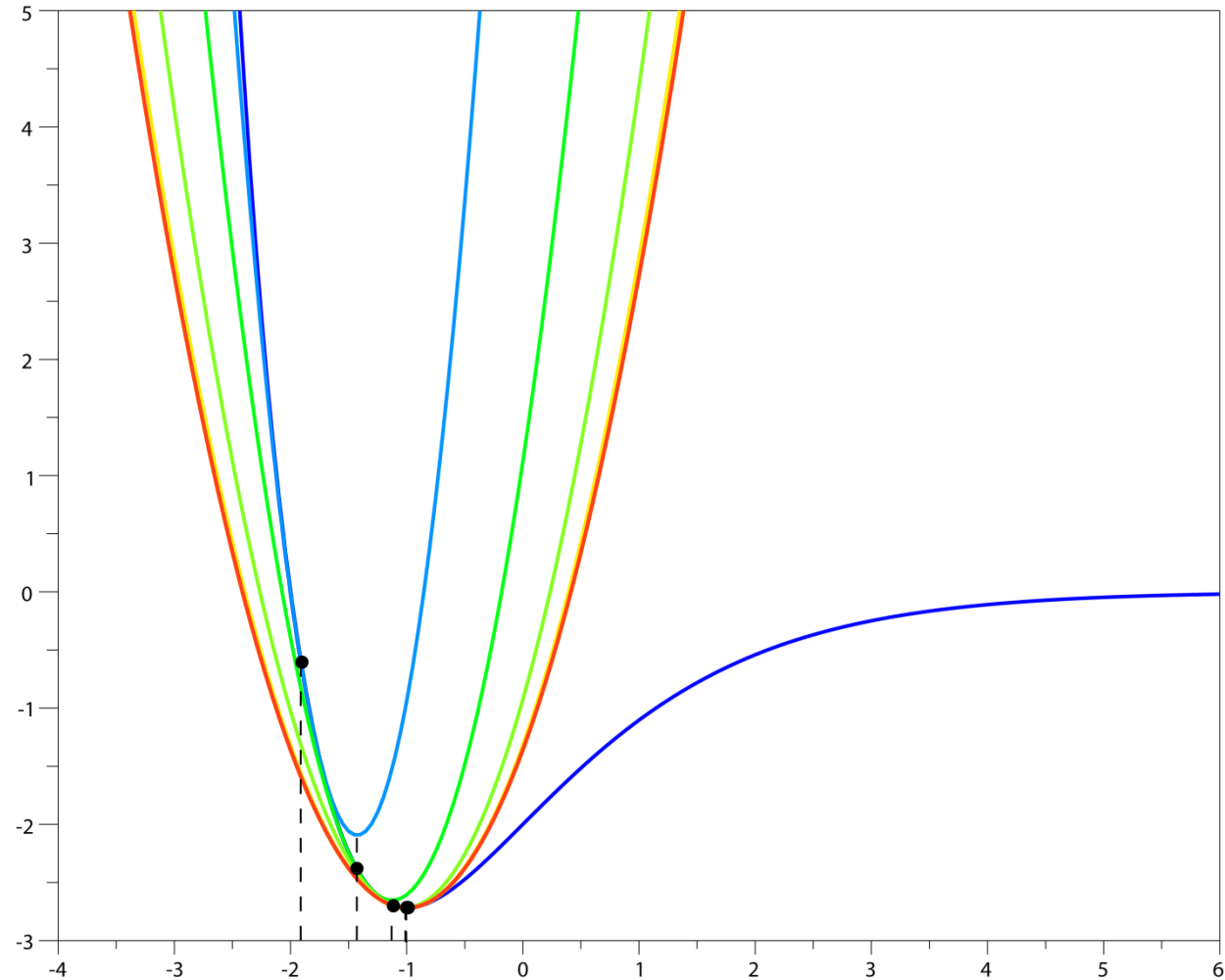
Problem specifications

- Problem specification
 - Computing $f(x)$ is easy
 - We can derivate $f: x \rightarrow f(x)$
 - We know the distance to the reference value



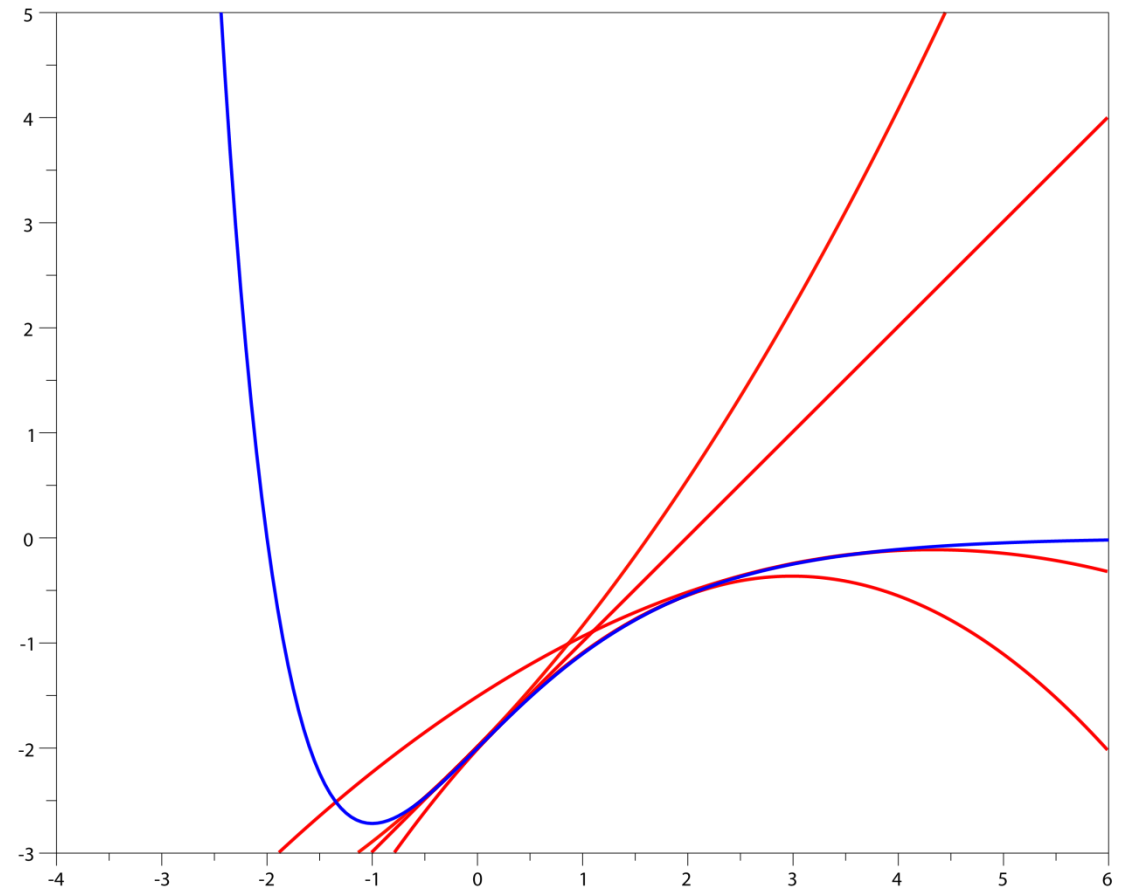
Newton method (unconstrained)

$x_0 = -1.9$
 $x_1 = -1.4263158$
 $x_2 = -1.1274228$
 $x_3 = -1.0144015$
 $x_4 = -1.0002045$
 $x_5 = -1.00000004$
 $x_6 = -1.$



Newton method (unconstrained)

- Ill-conditioned hessian
- Non positive hessian



Inverse geometry

- Decide: the robot configuration q
- Minimizing:

$$\|p(q) - p^*\|^2 \quad (3d \text{ objective})$$

$$\|\log({}^0M_i(q)^{-1} {}^0M_*)\|^2 \quad (6d \text{ objective})$$