

Cognitive Robotics

Motion (part 1)

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Rijeka 2018

Outline

Introduction

Kinematics of Poses

Kinematics of Drive Systems

Trajectories

Motion Planning

Motion Control

Motions of Legged Robots

Optimization/Learning of Motions

Biologically Inspired Motions

Motion

Motion:

Change of position(s) by certain actions/skills,
e.g. for locomotion or manipulation.

Great variety of natural and technical systems

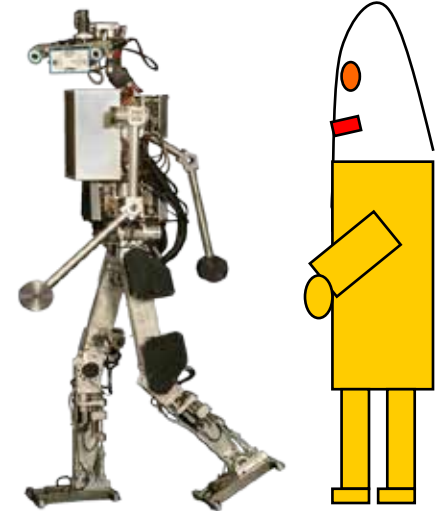
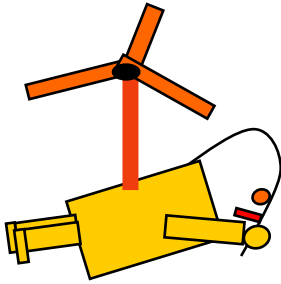
Formal description by mechanics

(force, mass, displacement, velocity, acceleration)

Problems in Robotics:

How can motions be realized and controlled
(hardware, software)

Locomotion (Ground, Air, Water, Space...)



Cars

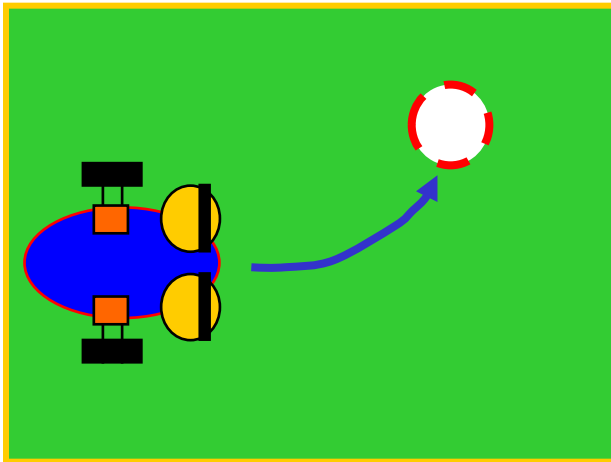


First Autonomous State Limousine

"MadeInGermany"

Autonomos Labs (R.Rojas, FU Berlin)

<https://www.youtube.com/watch?v=nX-le6JSU5g>



How many degrees of freedom?
Which poses can be reached?

Legged Robots, Special Designs

Hirose Robotics Lab, Tokyo



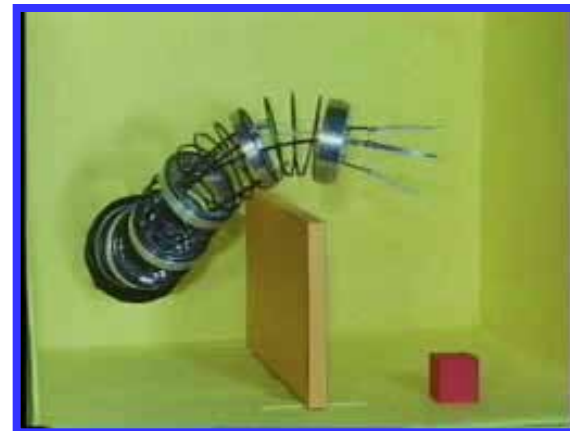
Rollerwalker



ACM-R5



Double Robotics



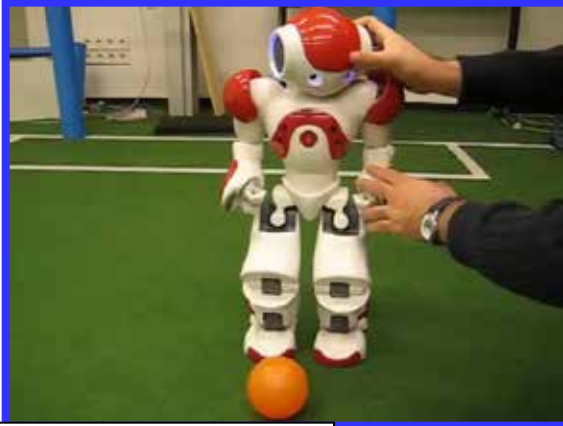
Active Endoscope "ELASTOR, Shape Memory Alloy Robot"

Boston Dynamics

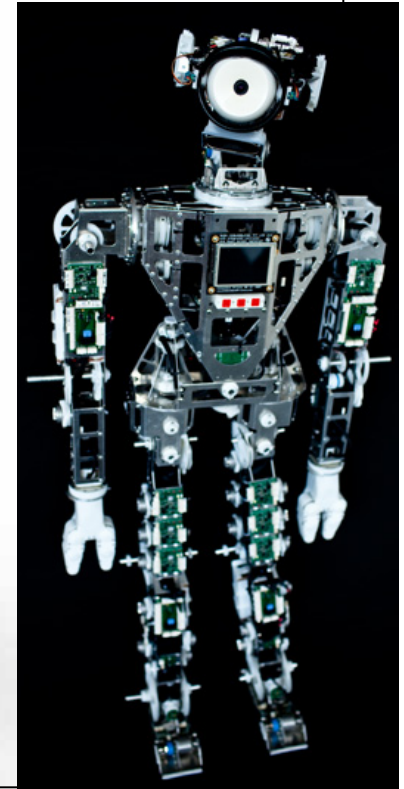
Sand Flee	Rhex
BigDog	RISE



Humanoid Robots



Nao (Aldebaran)



Myon
(Dr. Manfred Hild,
Neurorobotics Lab
Humboldt University)

How many degrees of freedom?
Which poses can be reached?

Flying Robots

Cognitive Robotics Lab
Prof. Verena Hafner, HU



Bionics

TU Berlin (Ingo Rechenberg)

<http://lautaro.bionik.tu-berlin.de/institut/s2foshow/>



Manipulators

KUKA Roboter GmbH



KUKA Roboter GmbH, Bachmann
WikiMedia

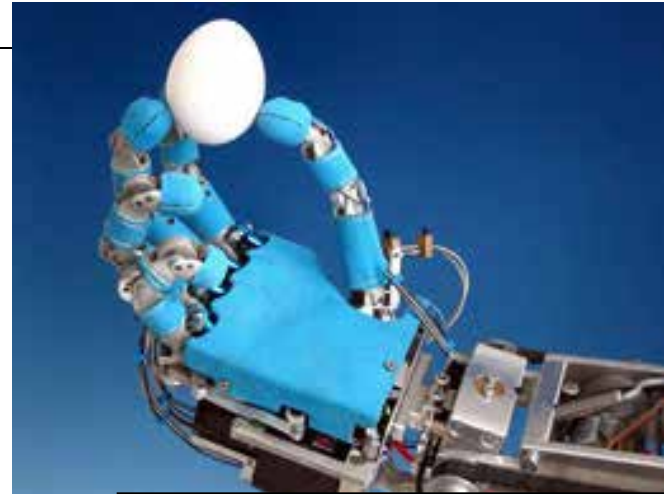
BMW Werk Leipzig
WikiMedia



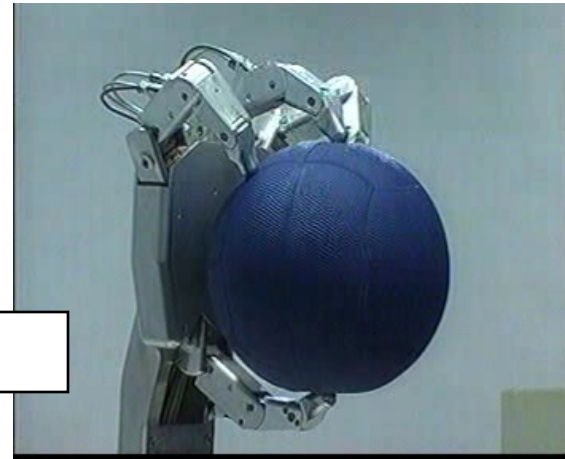
Manipulators



Karlsruhe



HU Berlin (Torsten Siedel)



Aachen

Manipulators



Surgical Robot DaVinci
Photo Nader Moussa
(WikiMedia)



Feeding Robot
Bestic AB (Stockholm)
Photo: Alice Öberg
(Sweden.se)

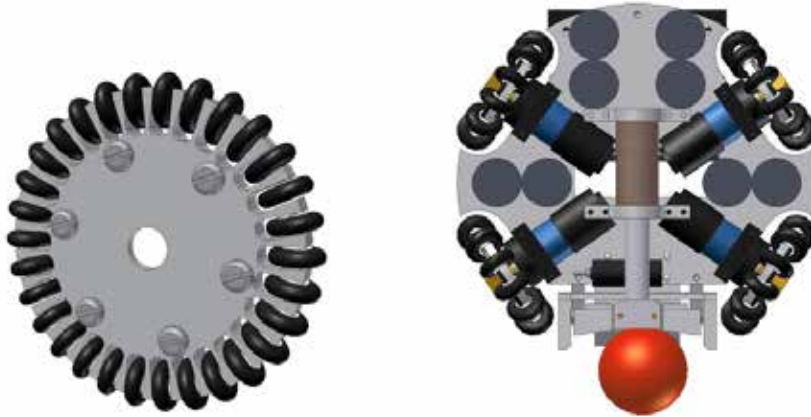
Manipulators



Photo: NASA
(WikiMedia)

Strawberry Harvesting Robot
by Robotic Harvesting LLC

Wheels, Chains, ...



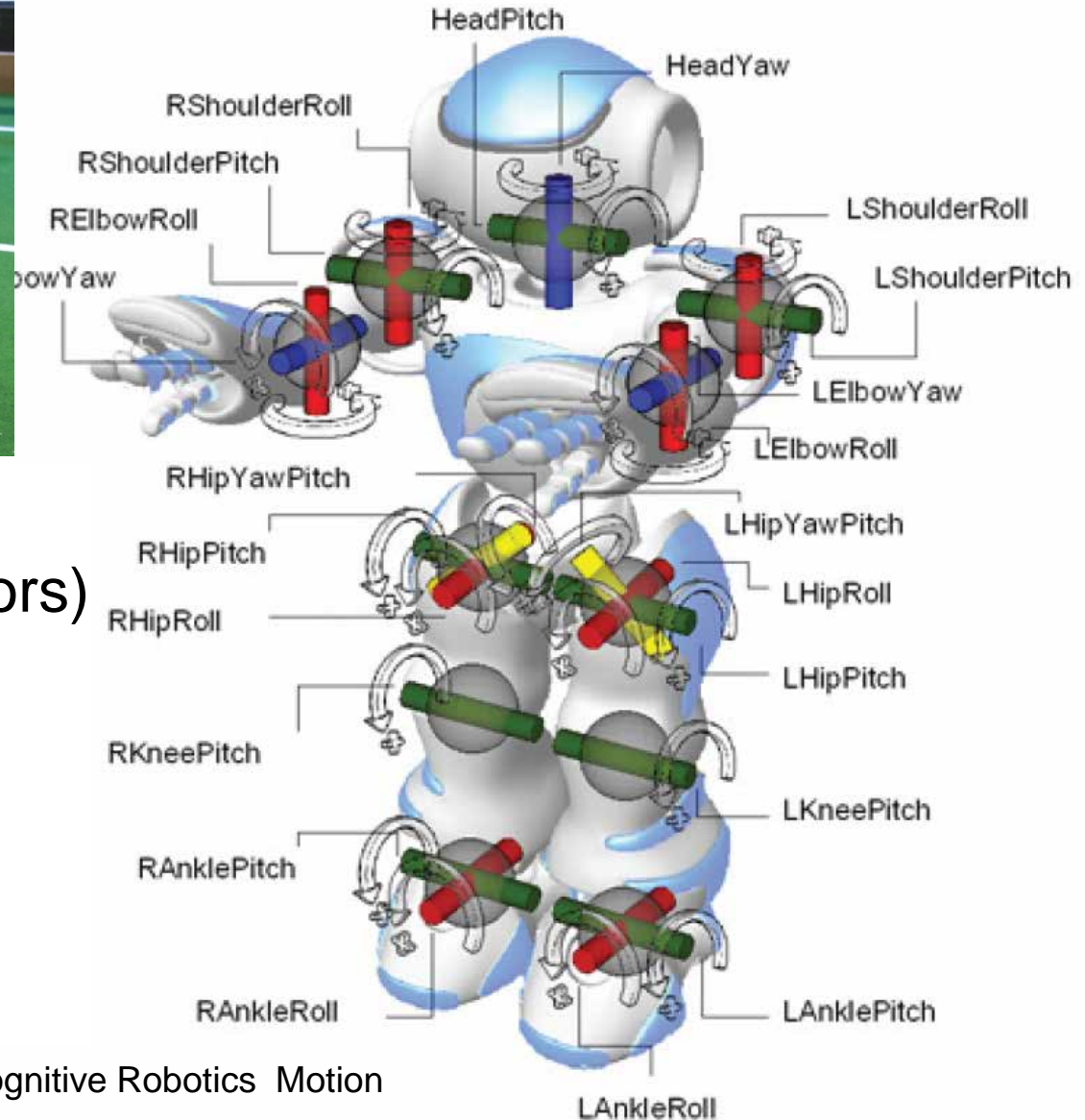
Joints

- Active: control with motors, pulleys, ...
- Problem: loading of gear axes
- Passive: Adaptation

Maintaining rest position by drives, gravity, friction, preload,...



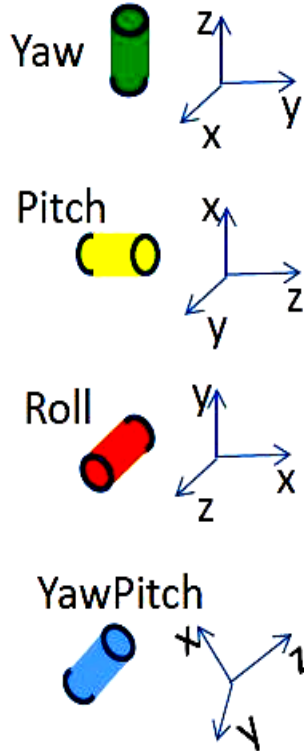
Joints of Nao from Aldebaran



21 active DOF (motors)

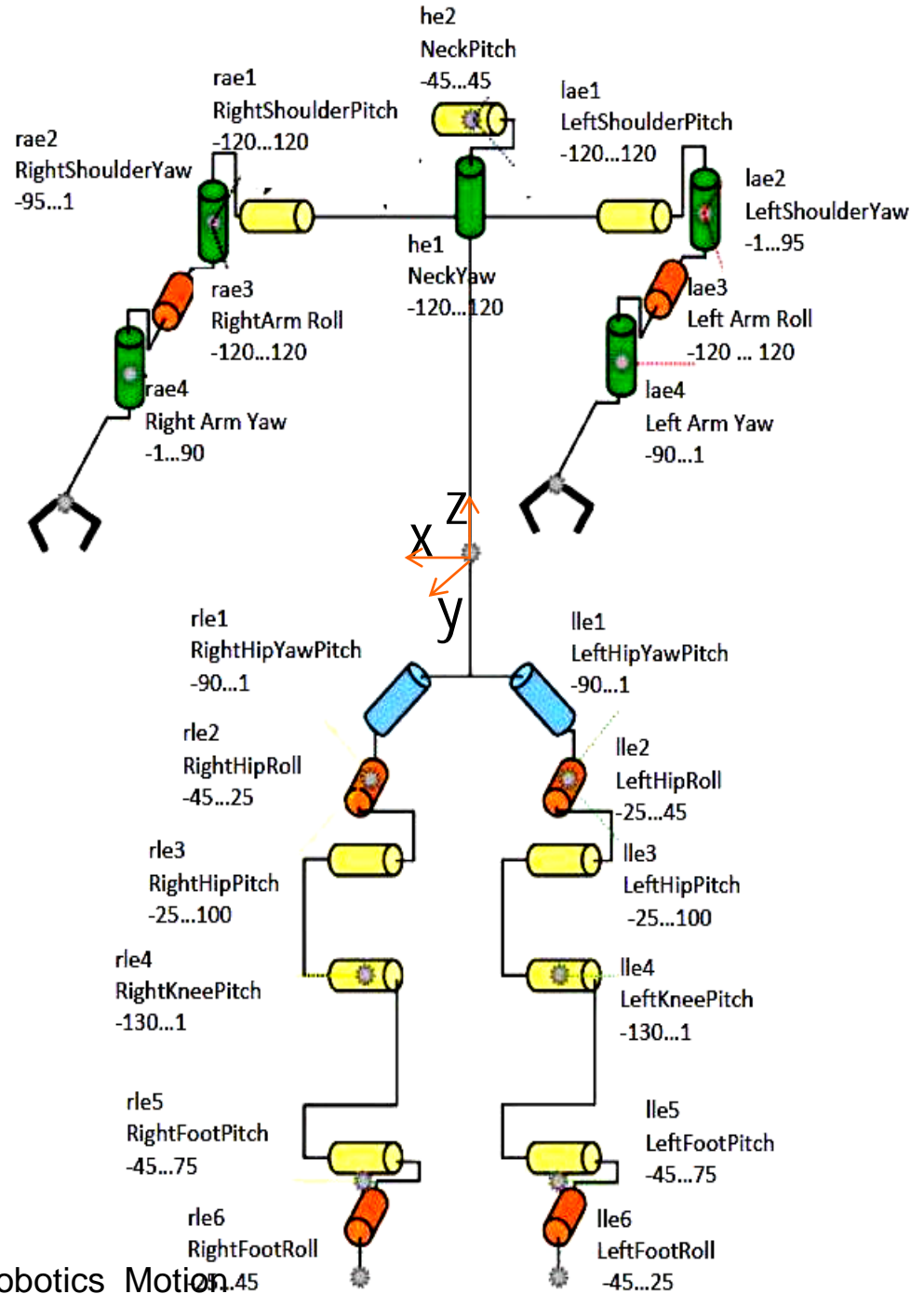
- 2 head
- 4 per arm
- 5 per leg
- 1 hip

Nao in Simulation



22 active DOF (motors):

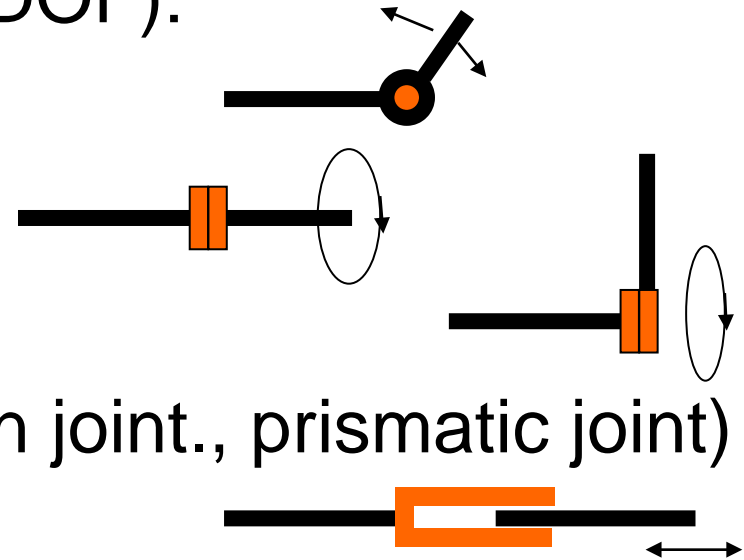
- 2 head
- 4 per arm
- 5 per leg
- 2 hip



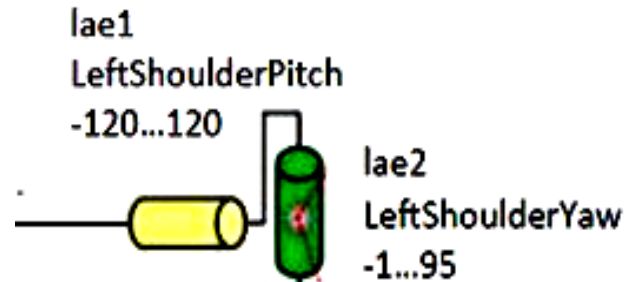
1 DOF Joints in Technique

With 1 degree of freedom (DOF):

- rotation joint
- torsion joint
- revolver joint
- Linear joint (Translation joint., prismatic joint)



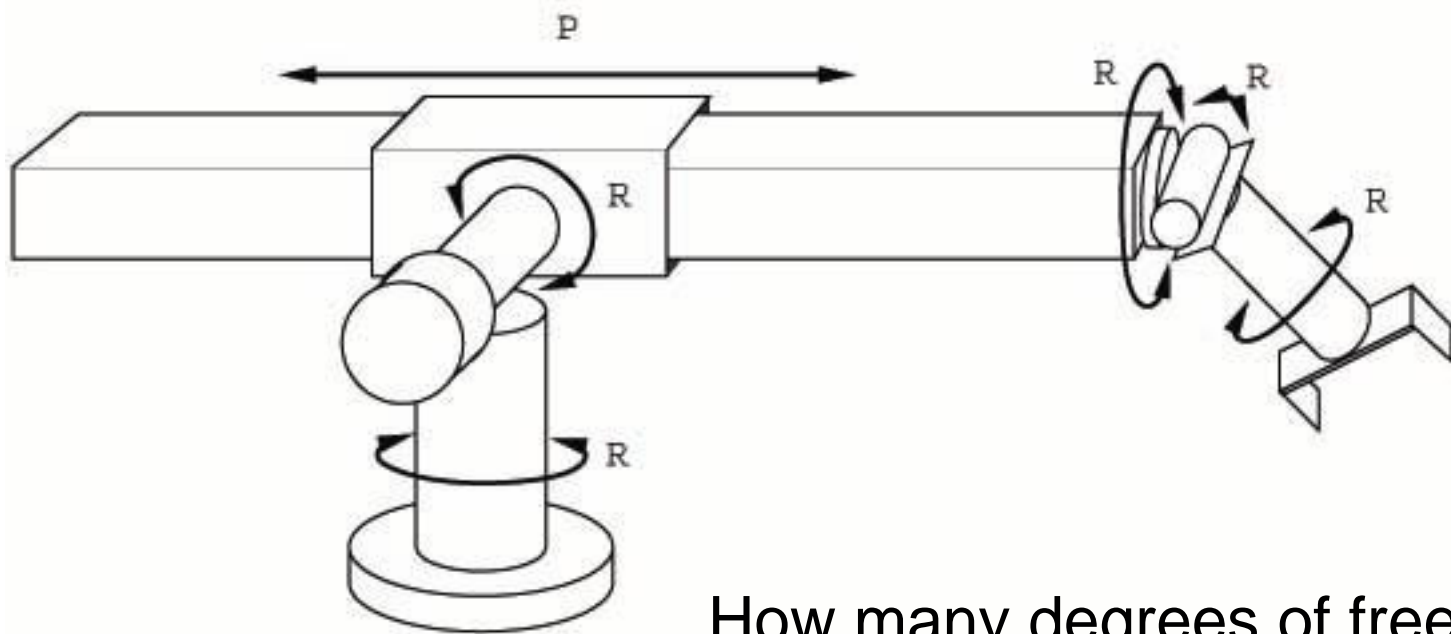
Several DOF by combinations



Stanford Manipulator

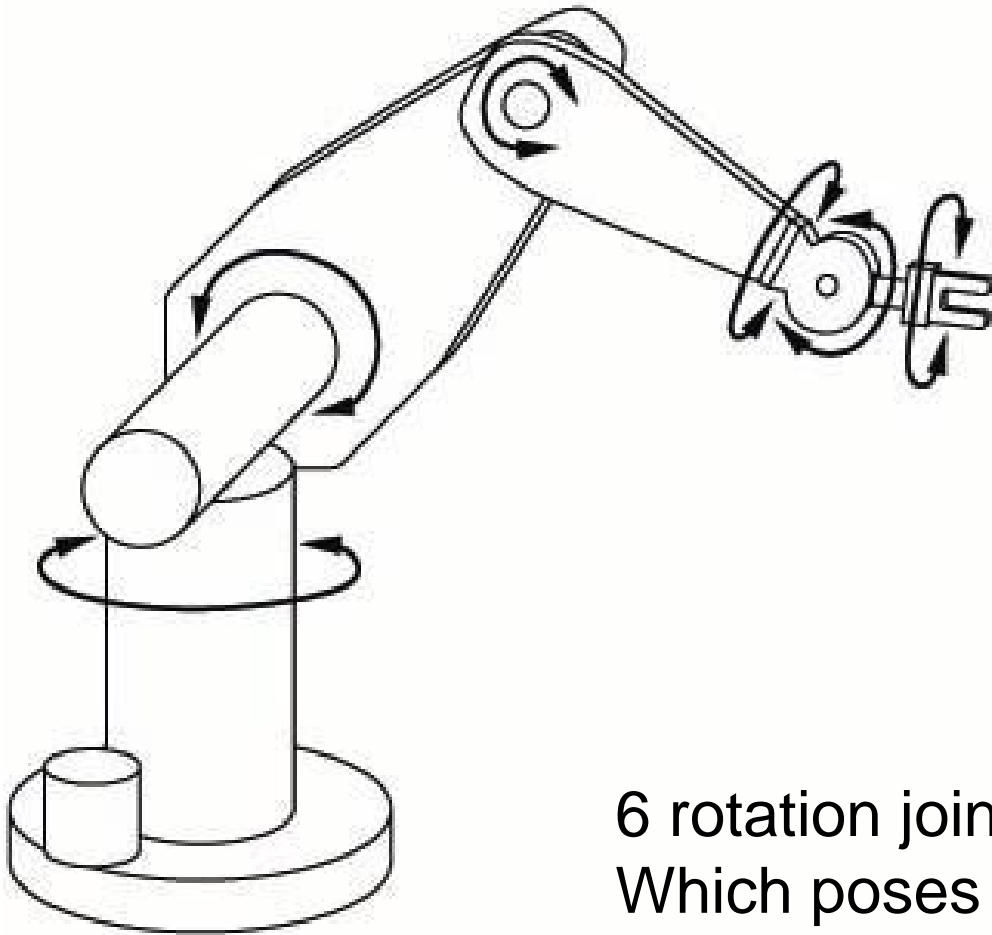
P: Prismatic joint

R: Rotation joint



How many degrees of freedom?
Which poses can be reached?

Puma (Programmable Universal Manipulation Arm)



6 rotation joints (6 DOF)
Which poses can be reached?

Degrees of Freedom (DOF)

DOF is the

- minimal number m of parameters p_1, \dots, p_m for complete description

equivalently:

- maximal number m of independent parameters p_1, \dots, p_m

Degrees of Freedom (DOF)

DOF of poses

(= parameters for complete description in work space):

- point on plan $p=(x,y)$, 2 DOF (2 position)
- car on plane: $p=(x,y,q)$, 3 DOF (2 position, 1 orientation)
- airplane: $p=(x,y,z, f, Y, q)$, 6 DOF (3 position, 3 orientation)

DOF of control parameters (in control/configuration space):
independently movable parts (joints, wheels/axes, ...)

DOF of control may be *active* (actuated) or *passiv*

Degrees of Freedom (DOF)

20 active DOF (motors)

- 3 per leg
- 3 head
- 2 tail
- 1 mouth
- 1 per ear



Degrees of Freedom (DOF)

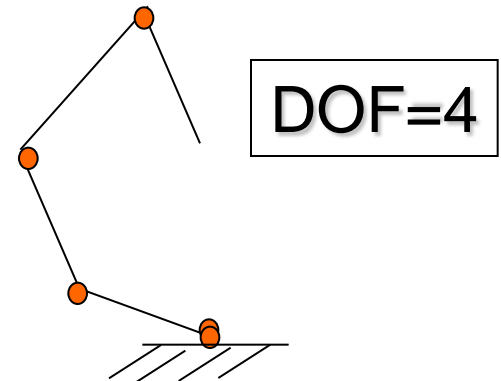
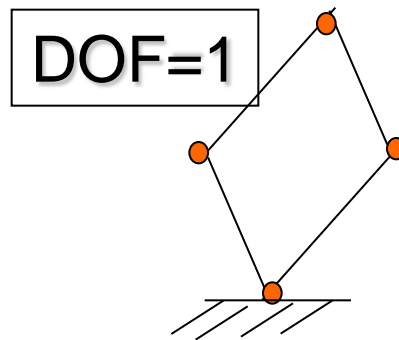


Degrees of Freedom (DOF)

Reachable poses depend on morphology and environment



Constraints $C(p_1, \dots, p_m) = 0$
for parameters
may reduce DOF



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Kinematics of Poses

Kinematics (forward kinematics):

- What is the pose?

Inverse kinematics (reverse kinematics):

- How to set the pose?

Simplification in Kinematics:
Neglect mass and force

Work Space and Configuration Space

Work space: „Relevant“ environment of the robot or some part.

Pose $p=(p_1, \dots, p_m)$:

Position/orientation of the robot or some part in work space (e.g. the pose of an end effector, of a camera etc.).

$m = \text{DOF of the pose in Workspace}$

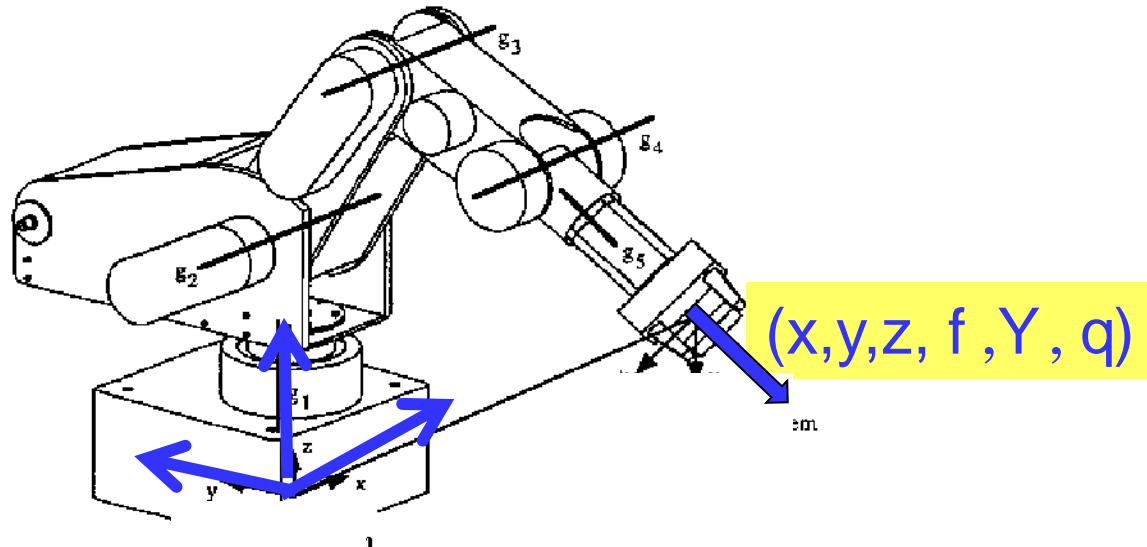
End effector of an industrial robot:

$p=(x,y,z, f, Y, q)$

6 DOF:

Ø 3 position,

Ø 3 orientation

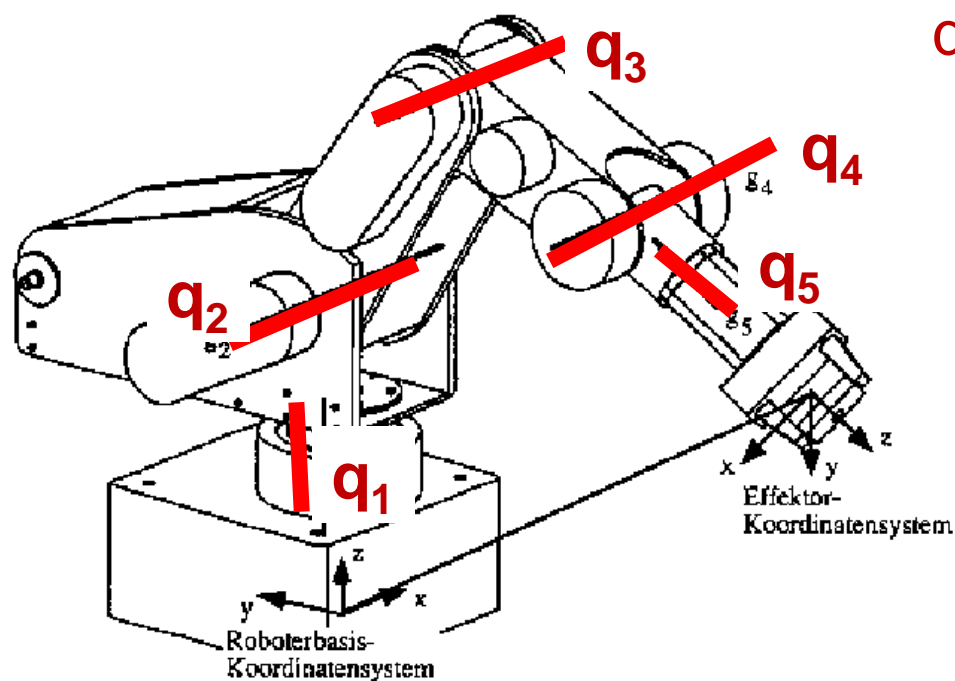


Work Space and Configuration Space

Configuration space:

Configuration $q=(q_1, \dots, q_n)$: parameters of joints etc.
„generalized coordinates“, „control parameters“

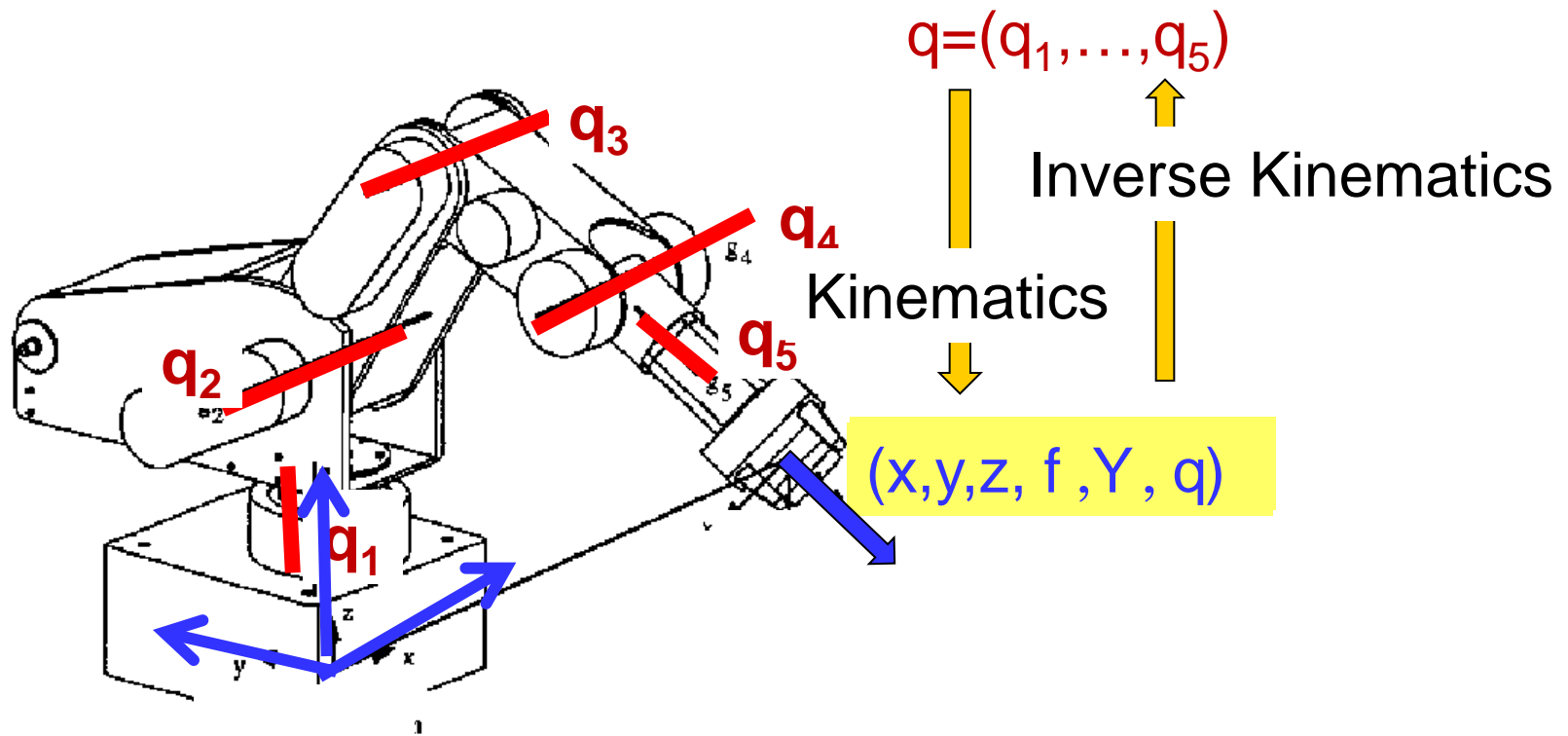
$n = \text{DOF in configuration space}$



$q=(q_1, \dots, q_5)$

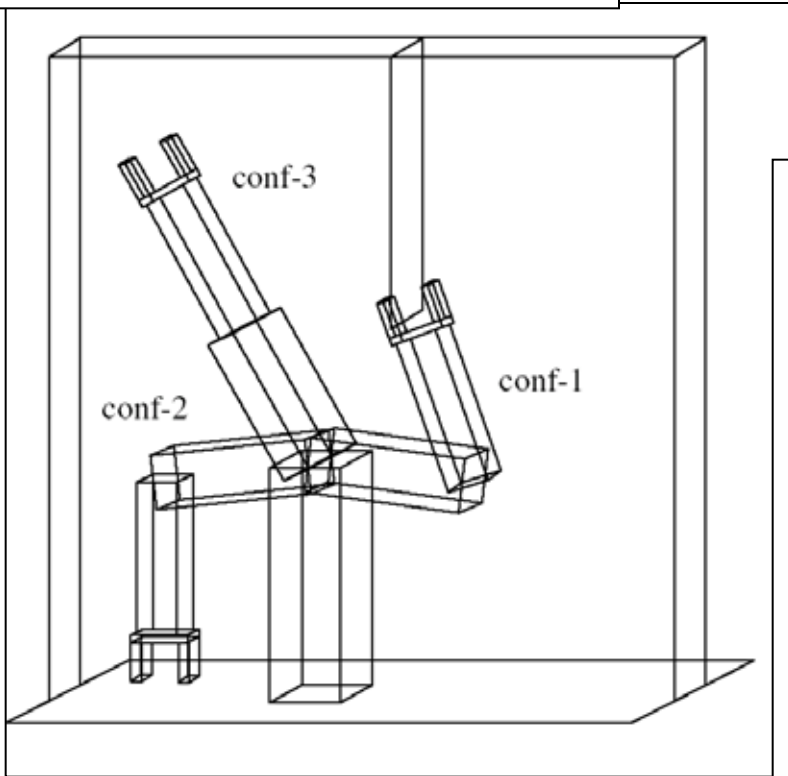
5 DOF

Work Space and Configuration Space



Work Space and Configuration Space

Work Space

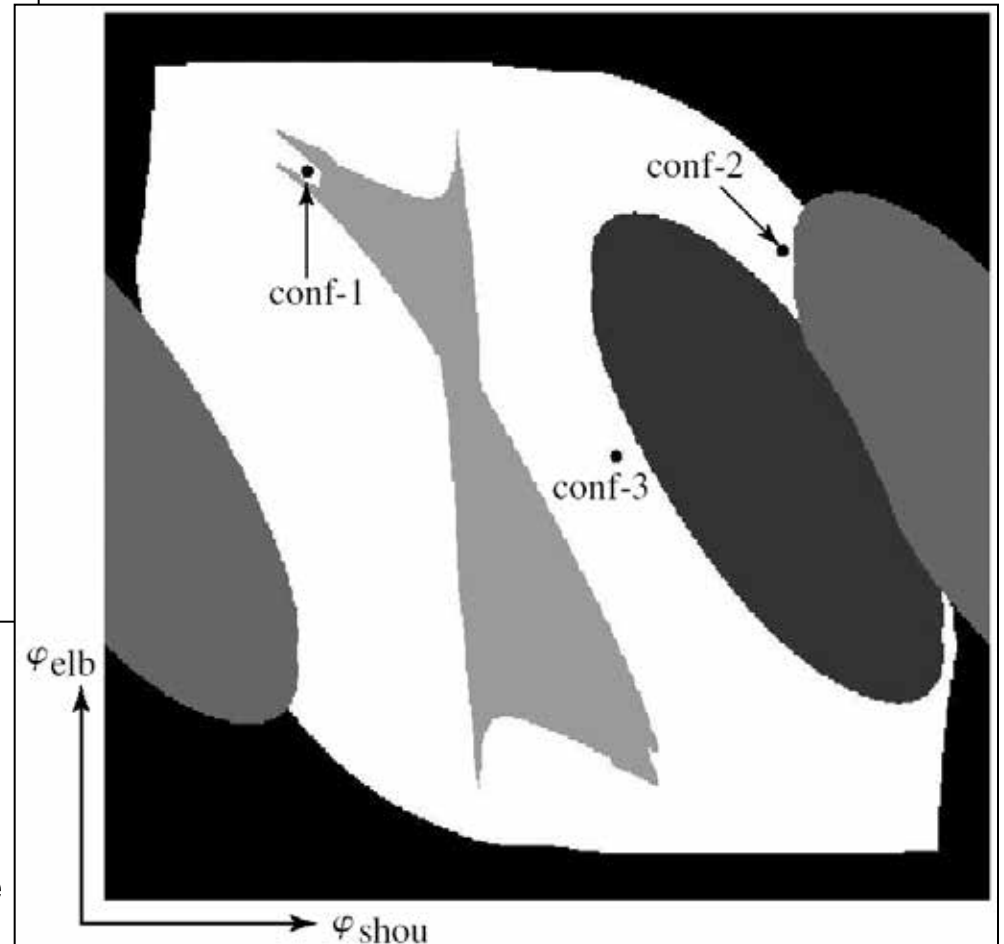


From Russell/Norvig:
Artificial Intelligence

Burkhard

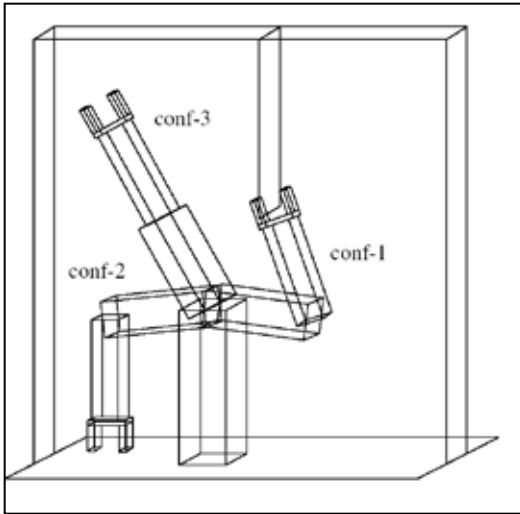
Cognitive

Configuration Space

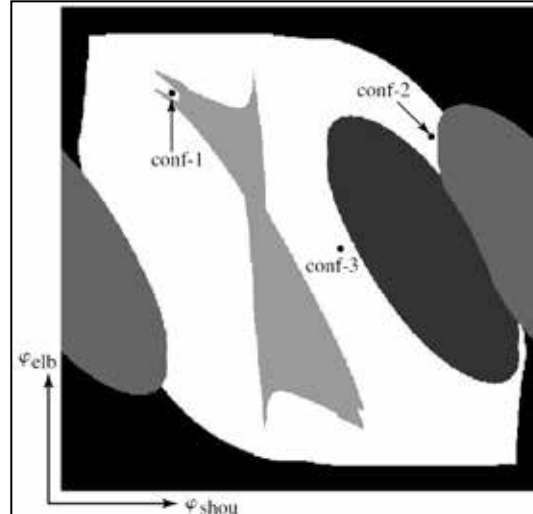


Constraints

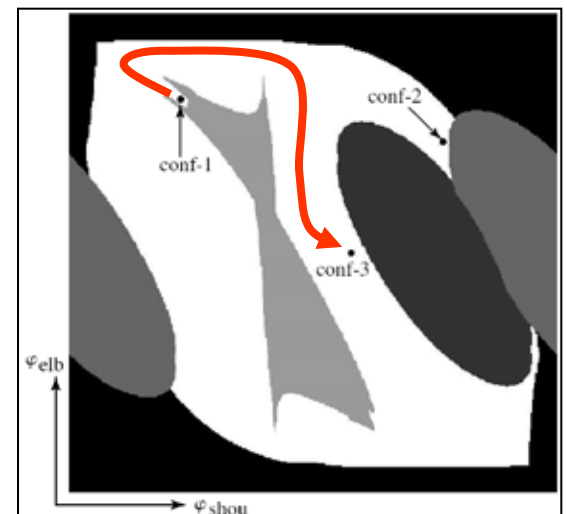
In Work Space
by morphology
and
environment



In Configuration
Space by related
unreachable
region



Motion planning
in Configuration
Space



Work Space and Configuration Space

Kinematics:

$$p = f(q)$$

Determine pose from configuration

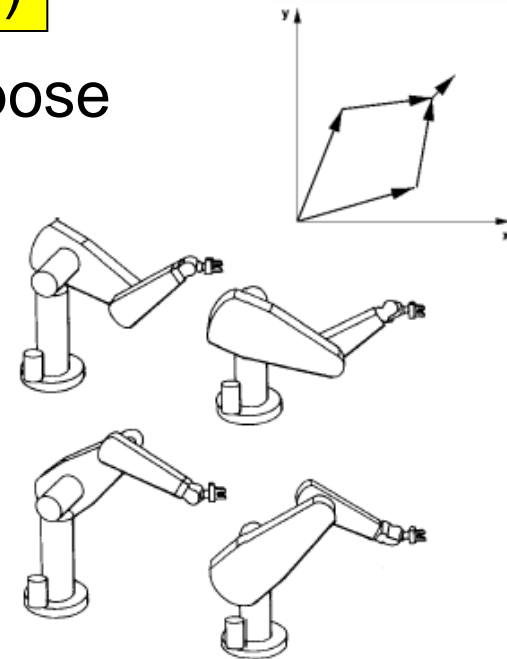
- Configuration determines pose uniquely

Inverse Kinematics:

$$q = f^{-1}(p)$$

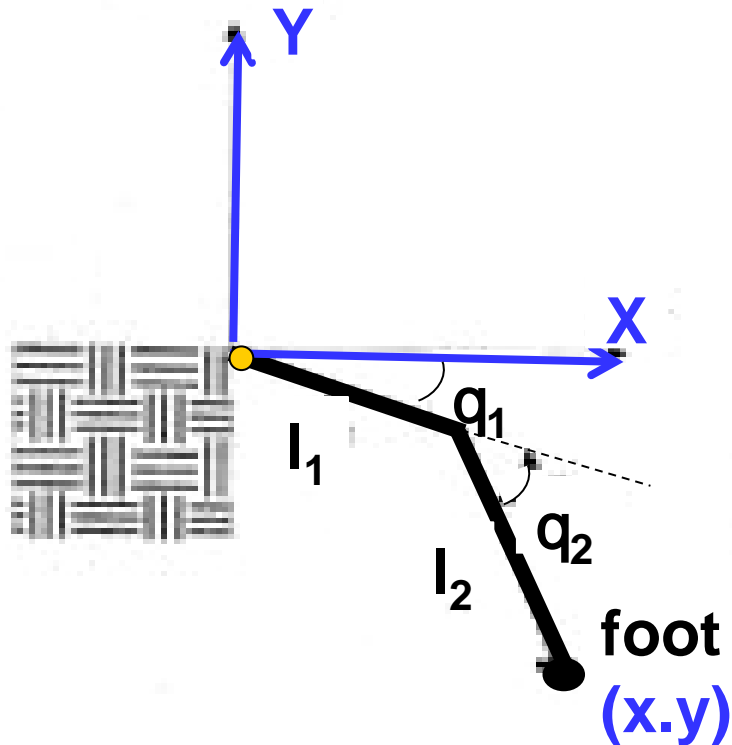
Find a configuration for requested pose

- Pose might be realized by different configurations

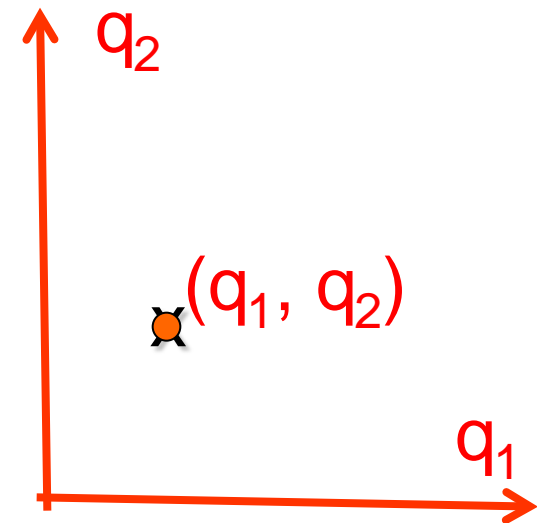


Example „Planar Leg“

Work space x, y

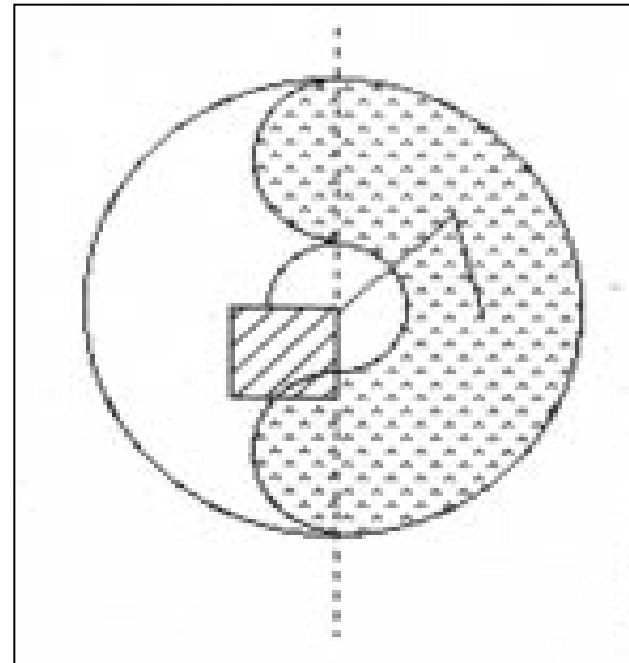
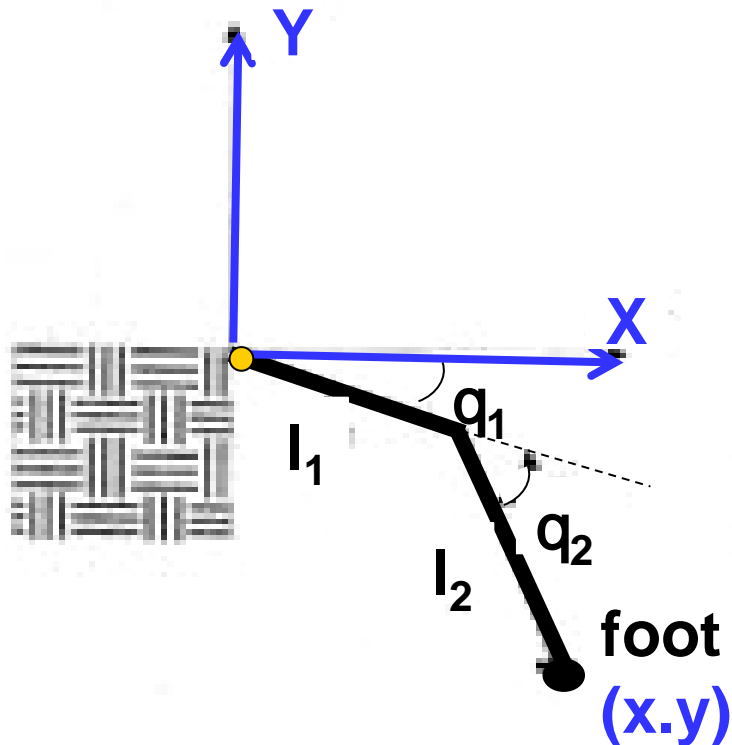


Configuration space q_1, q_2



Example „Planar Leg“

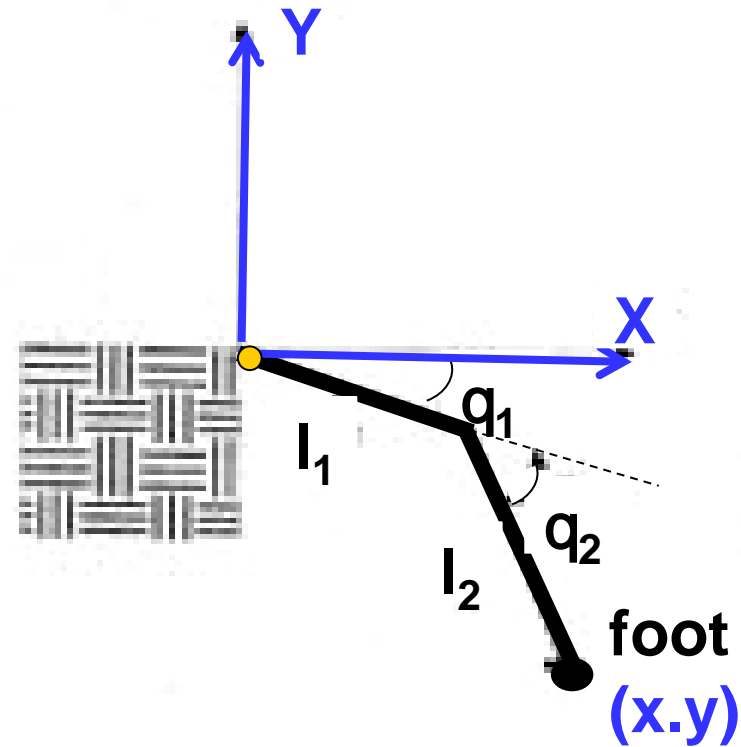
Achievable points are limited by joint angles q_1 , q_2 , limb lengths l_1 , l_2 , spacial constraints



Example „Planar Leg“

Kinematics:

- Rotation by Q_1
- Translation by l_1
- Rotation by Q_2
- Translation by l_2



$$\begin{bmatrix} x \\ y \end{bmatrix} = l_1 \begin{bmatrix} \cos(\theta_1) \\ \sin(\theta_1) \end{bmatrix} + l_2 \begin{bmatrix} \cos(\theta_1 + \theta_2) \\ \sin(\theta_1 + \theta_2) \end{bmatrix}$$

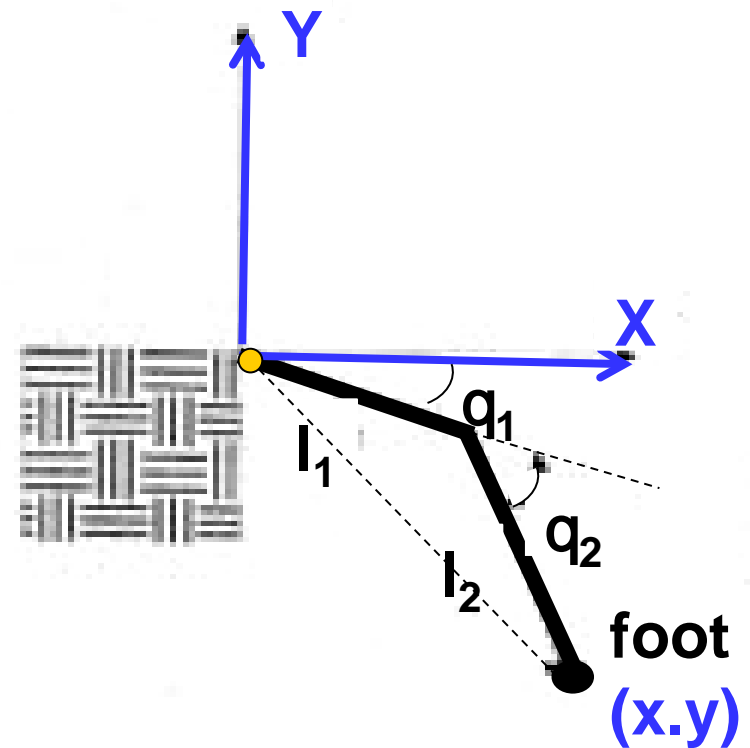
Example „Planar Leg“

Inverse Kinematics:

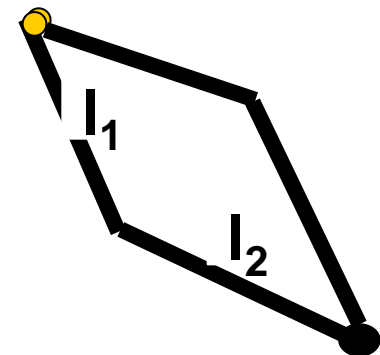
(by cosine rule)

$$\cos(\theta_2) = \frac{x^2 + y^2 - l_1^2 - l_2^2}{2l_1l_2}$$

$\cos(Q_1)$ computable by the
formula for forward kinematics



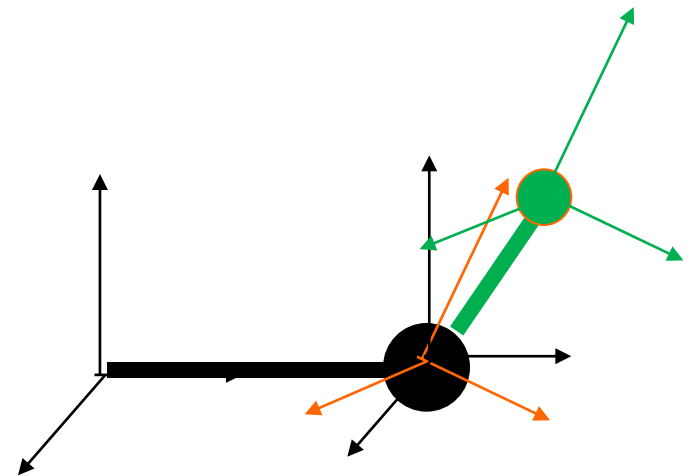
2 possible solutions:



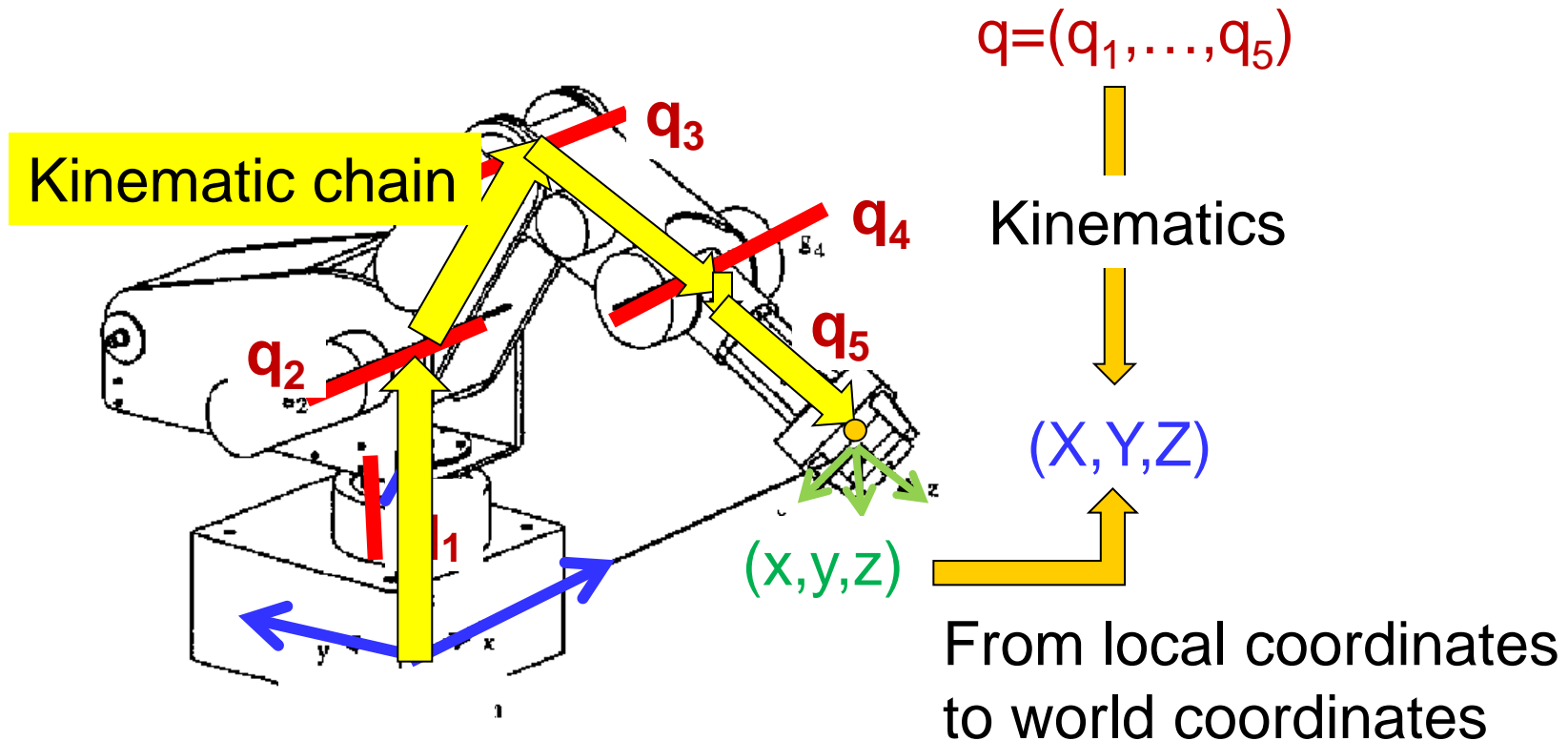
Kinematics: Calculate $p = f(q)$

Joints: Rotations

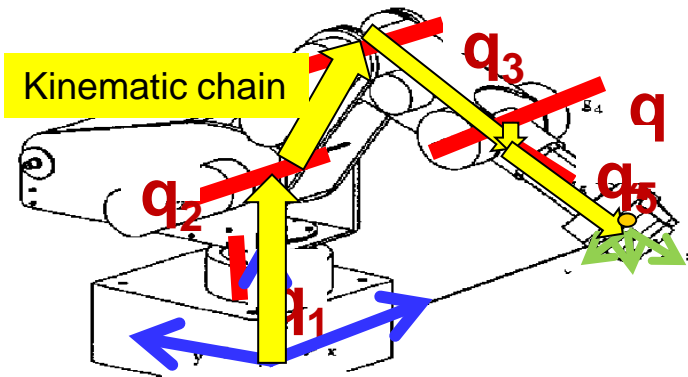
Limbs: Translations



Kinematics: Coordinate Transformation



Kinematics: Coordinate Transformation



Coordinate transformation by a sequence of intrinsic rotations and translations along the **cinematic chain**.

The ordering must be preserved.

Homogenous Coordinates for 3D

4-dimensional vector

($x/w, y/w, z/w, w$) with arbitrary $w \neq 0$ represents (x,y,z)

We will use $(x, y, z, 1)$ i.e. $w=1$

The 4-dimensional matrix H
can describe Rotation R
followed by Translation T

$$H = \begin{bmatrix} R & T \\ 0 & 1 \end{bmatrix}$$

$$H \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} R \begin{bmatrix} x \\ y \\ z \end{bmatrix} \\ 1 \end{bmatrix} + \begin{bmatrix} T \\ 1 \end{bmatrix}$$

Homogenous Coordinates for 3D

Sequence of transformations along cinematic chain
can be described by matrix multiplications

$$M = H_1 \times H_2 \times H_3 \times \dots \times H_n$$

$$X = M \times x$$
$$\begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} = M \times \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

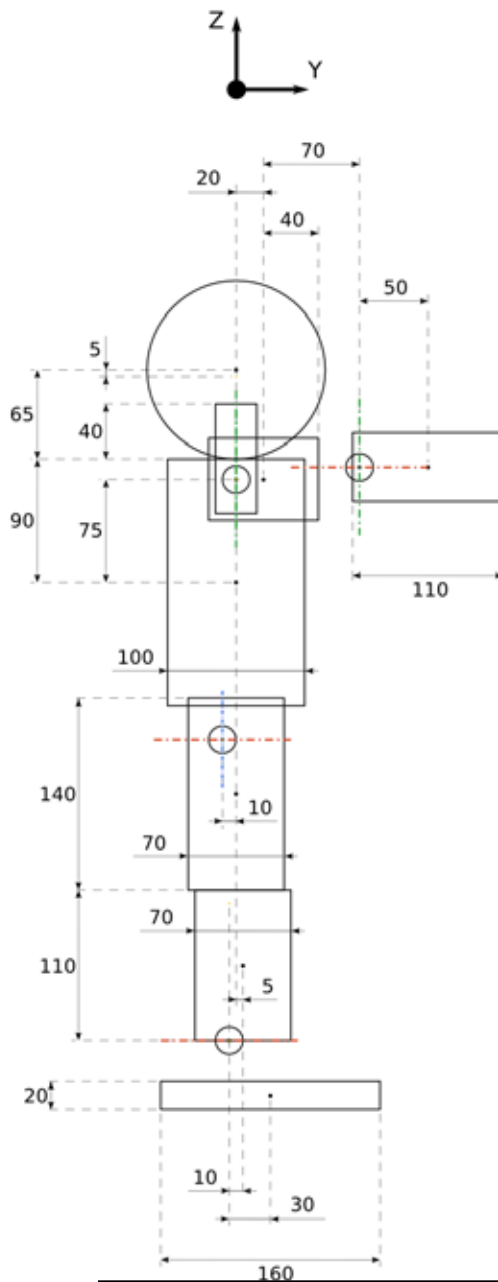
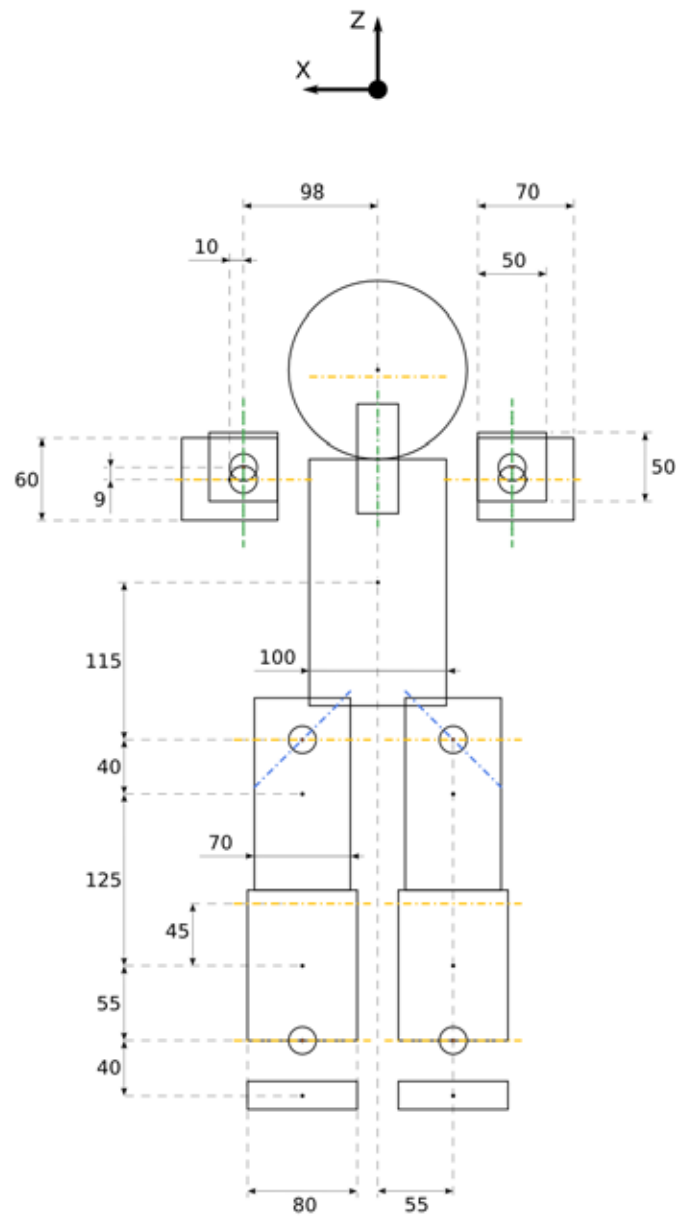
- **Kinematics**

by computing X from $X = M \times x$ for given M , x

- **Inverse Kinematics**

by finding $M = H_1 \times H_2 \times H_3 \times \dots \times H_n$ for given X , x

(but usually by other calculations resp. approximations)



Nao robot model (All values in mm)	
Total dimensions (x, y, z):	286, 260, 605 mm
----- Pitch	- - - - - Yaw
----- Roll	- - - - - YawPitch

Kinematics need information on Morphology:

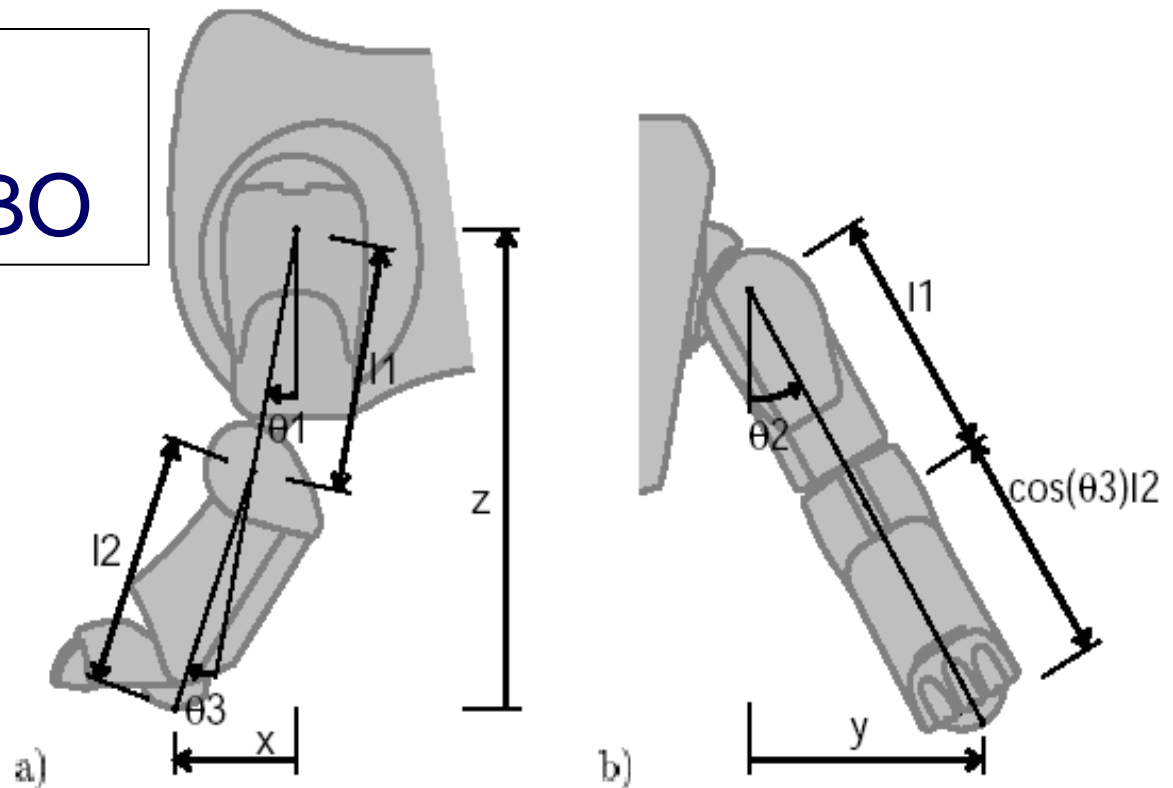
Body parts and joints of Simulated Nao

Body part					Hinge Joint				
Name	Parent	Translation	Mass	Geometry	Name	Anchor	Axis	Min	Max
torso			1.2171	Box 0.100, 0.100, 0.180					
neck	torso	0, 0, 0.090	0.05	Cylinder L: 0.080 R: 0.015	HJ1	0, 0, 0	0,0,1	-120	120
head	neck	0, 0, 0.065	0.35	Sphere 0.065	HJ2	0, 0, -0.005	1,0,0	-45	45
shoulder	torso	0.098, 0, 0.075(r) -0.098, 0, 0.075(l)	0.07	Sphere 0.010	AJ1	0, 0, 0	1,0,0	-120	120
upperarm	shoulder	0.010, 0.020, 0(r) -0.010, 0.020, 0(l)	0.15	Box 0.07, 0.08, 0.06	AJ2	-Translation	0,0,1	-95(r) -1(l)	1(r) 95(l)
elbow	upperarm	-0.010, 0.070, 0.009(r) 0.010, 0.070, 0.009(l)	0.035	Sphere 0.010	AJ3	0, 0, 0	0,1,0	-120	120
lowerarm	elbow	0, 0.050, 0	0.2	Box 0.050, 0.110, 0.050	AJ4	-Translation	0,0,1	-1(r) -90(l)	90(r) 1(l)
hip1	torso	0.055, -0.010, -0.115(r) -0.055, -0.010, -0.115(l)	0.09	Sphere 0.010	LJ1	0, 0, 0	-0.7071, 0, 0.7071(r) -0.7071, 0, -0.7071(l)	-90	1
hip2	hip1	0, 0, 0	0.125	Sphere 0.010	LJ2	0, 0, 0	0,1,0	-45(r) -25(l)	25(r) 45(l)
thigh	hip2	0, 0.010, -0.040	0.275	Box 0.070, 0.070, 0.140	LJ3	-Translation	1,0,0	-25	100
shank	thigh	0, 0.005, -0.125	0.225	Box 0.080, 0.070, 0.110	LJ4	0,-0.010, 0.045	1,0,0	-130	1
ankle	shank	0, -0.010, -0.055	0.125	Sphere 0.010	LJ5	0, 0, 0	1,0,0	-45	75
foot	ankle	0, 0.030, -0.040	0.2	Box 0.080, 0.160, 0.020	LJ6	-Translation	0,1,0	-25(r) -45(l)	45(r) 25(l)

- Name is the body part name of Nao
 - Parent is the parent of the body
 - Translation is the offset relative to its parent (in meter)
 - Mass is the mass of this body (in kilogram)
 - Geometry is the size of its geometry representation (in meter)
 - Name is the joint name installed on this body
 - Anchor is the offset of the joint anchor relative to the body that installed on in meter
 - Axis is the joint axis relative to the body that installed on (x, y, z-orientation of the axis)
 - Min is the min angle that the joint can reach (in degrees)
 - Max is the max angle that the joint can reach (in degrees)
- Note: All values are relative to the torso coordinate system!
(which faces the y-axis)

Example: Kinematics AIBO

Diploma thesis
Uwe Düffert



World coordinates in the shoulder.

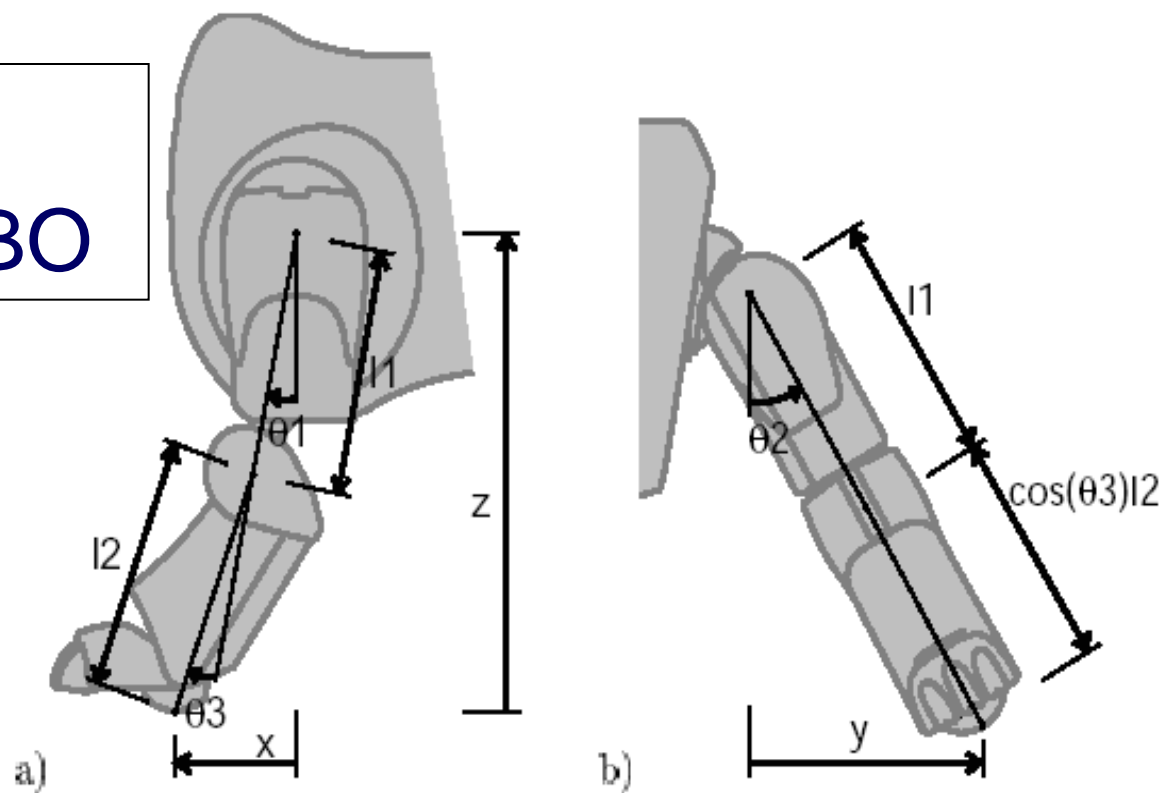
What are the coordinates (x,y,z) of the left forefoot?

Calculation:

by transformation of the foot coordinates to shoulder coordinates

Example: Kinematics AIBO

Diploma thesis
Uwe Duffert

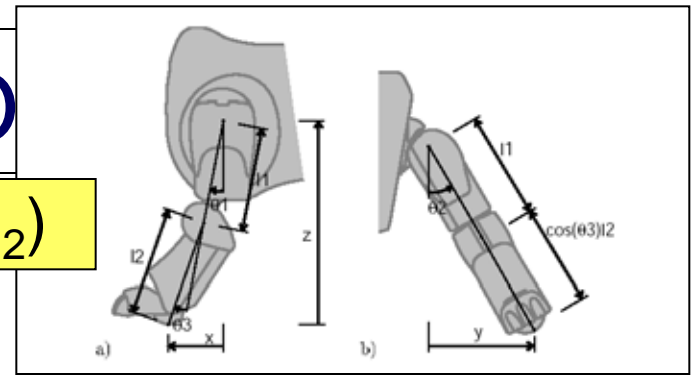


- Transformation of the foot coordinates to shoulder coordinates:
1. Translation lower leg: shift towards negative z axis (l_2).
 2. Rotation knee: rotate clockwise around y -axis (q_3).
 3. Translation upper leg: shift towards negative z axis (l_1).
 4. Rotation shoulder 2: rotate counter-clockw. around x -axis (q_2).
 5. Rotation shoulder 1: rotate clockwise around y -axis (q_1).

$$\text{Rot}(-q_1) \text{Rot}(q_2) \text{Trans}(l_1) \text{Rot}(-q_3) \text{Trans}(l_2)$$

Example: Kinematics AIBO

Rot(-q₁) Rot(q₂) Trans(l₁) Rot(-q₃) Trans(l₂)



$$\begin{aligned}
 \begin{pmatrix} x \\ y \\ z \\ 1 \end{pmatrix} &= \text{Rot}_y(-\theta_1) \cdot \text{Rot}_x(\theta_2) \cdot \text{Trans} \begin{pmatrix} 0 \\ 0 \\ -l_1 \end{pmatrix} \cdot \text{Rot}_y(-\theta_3) \cdot \text{Trans} \begin{pmatrix} 0 \\ 0 \\ -l_2 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \\
 &= \begin{pmatrix} \cos(\theta_1) & 0 & -\sin(\theta_1) & 0 \\ 0 & 1 & 0 & 0 \\ \sin(\theta_1) & 0 & \cos(\theta_1) & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos(\theta_2) & -\sin(\theta_2) & 0 \\ 0 & \sin(\theta_2) & \cos(\theta_2) & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \cdot \\
 &\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & -l_1 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \cos(\theta_3) & 0 & -\sin(\theta_3) & 0 \\ 0 & 1 & 0 & 0 \\ \sin(\theta_3) & 0 & \cos(\theta_3) & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & -l_2 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \\
 &= \begin{pmatrix} l_2 \cos(\theta_1) \sin(\theta_3) + l_2 \sin(\theta_1) \cos(\theta_2) \cos(\theta_3) + l_1 \sin(\theta_1) \cos(\theta_2) \\ l_1 \sin(\theta_2) + l_2 \sin(\theta_2) \cos(\theta_3) \\ l_2 \sin(\theta_1) \sin(\theta_3) - l_2 \cos(\theta_1) \cos(\theta_2) \cos(\theta_3) - l_1 \cos(\theta_1) \cos(\theta_2) \\ 1 \end{pmatrix} \cdot
 \end{aligned}$$

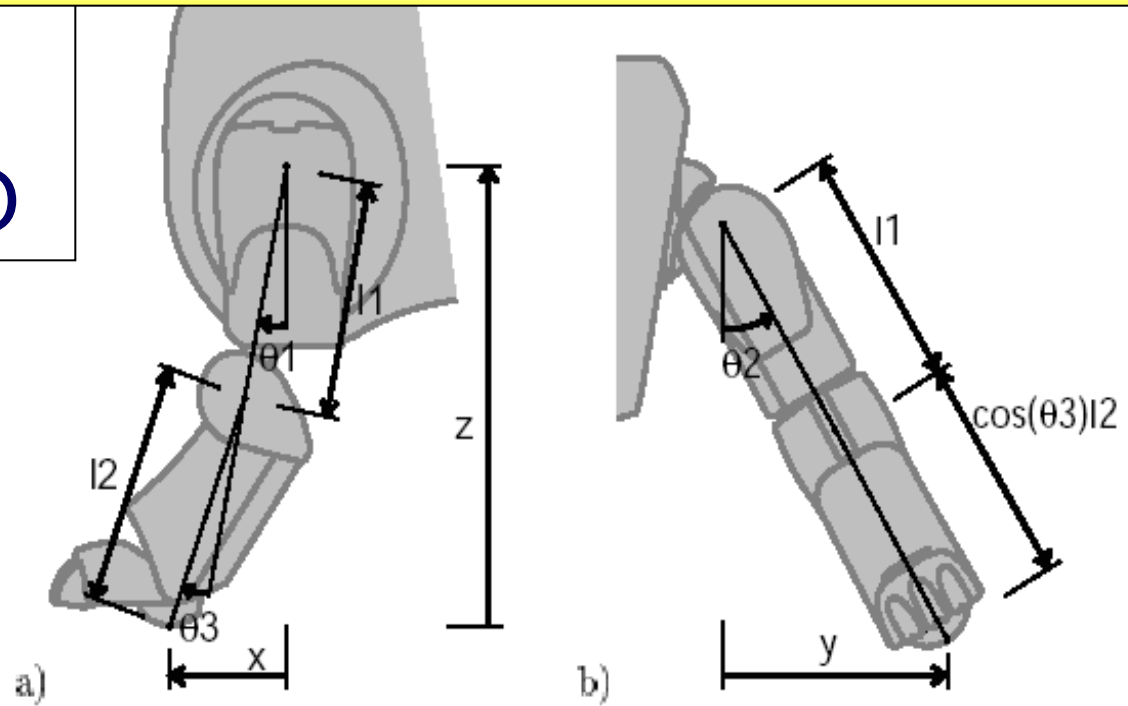
Example:
Inverse
Kinematics AIBO

Calculate q_1, q_2, q_3 for feet position (x, y, z)

q_3 (between l_1, l_2)
 by Cosine rule.

Preferably bending
 forward:

Positive solution.



$$\cos(\pi - \theta_3) = \frac{l_1^2 + l_2^2 - (x^2 + y^2 + z^2)}{2l_1l_2}$$

$$\theta_3 = \pi \pm \arccos\left(\frac{l_1^2 + l_2^2 - (x^2 + y^2 + z^2)}{2l_1l_2}\right)$$

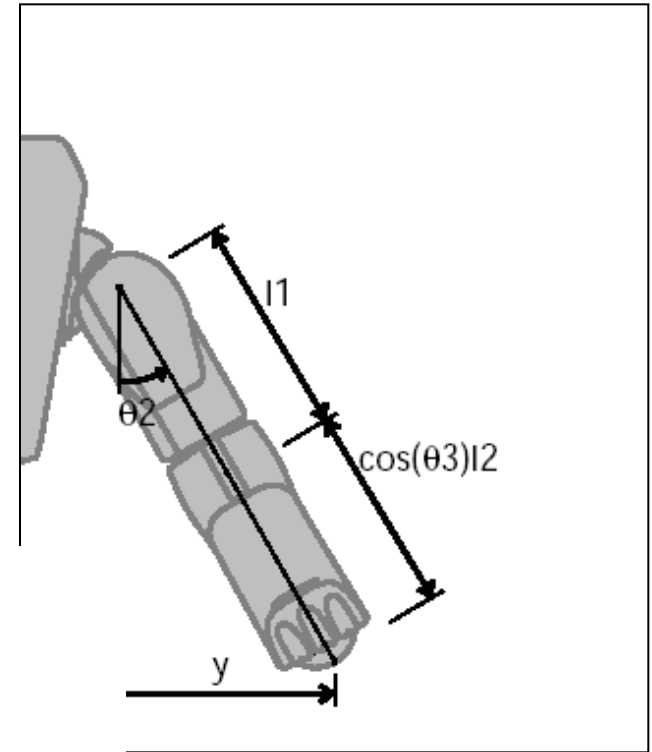
$$= \mp \arccos\left(\frac{(x^2 + y^2 + z^2) - l_1^2 - l_2^2}{2l_1l_2}\right)$$

Example: Inverse Kinematics AIBO

q_2 by definition of Sine,
where $|q_2| \leq \pi/2$ by
anatomy

$$y = \sin(\theta_2) \cdot (l_1 + l_2 \cdot \cos(\theta_3))$$

$$\theta_2 = \arcsin \left(\frac{y}{l_1 + l_2 \cos(\theta_3)} \right)$$



Example: Inverse Kinematics AIBO

$$a = l_2 \sin(\theta_3)$$

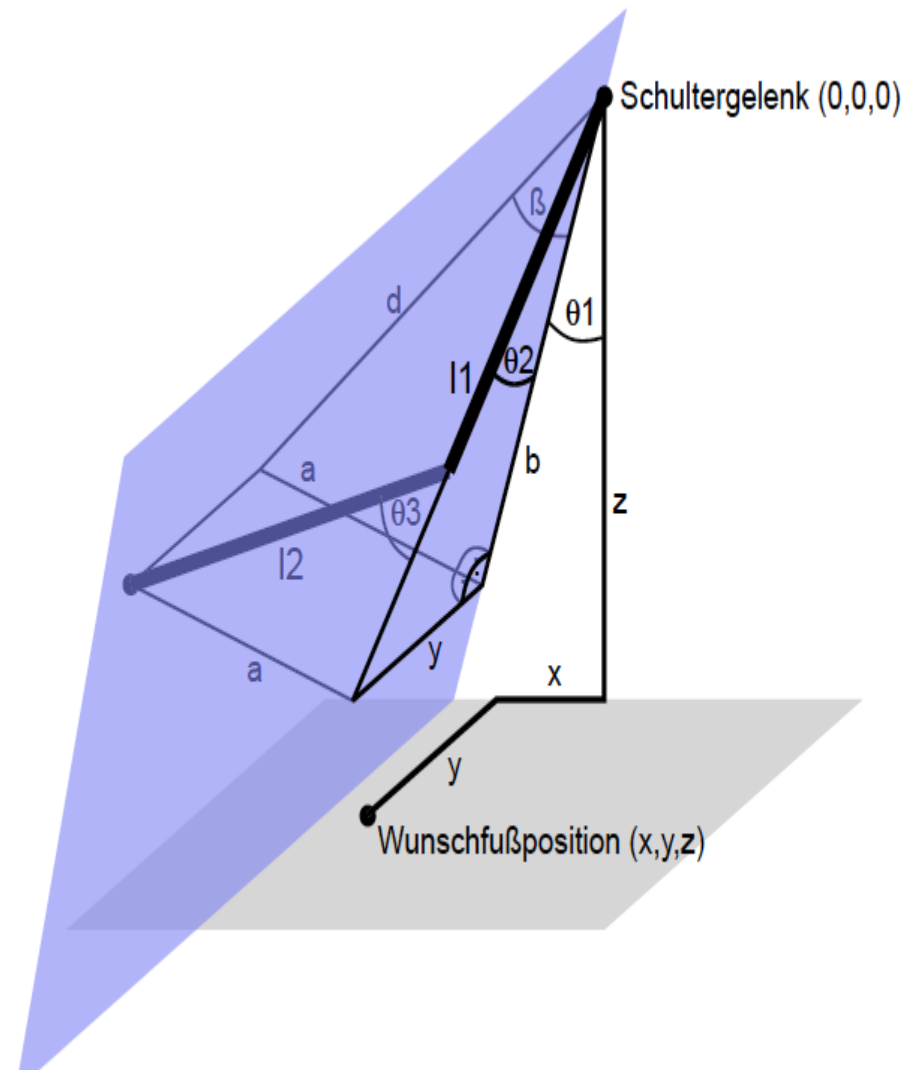
$$b = (l_1 + l_2 \cos(\theta_3)) \cos(\theta_2)$$

$$d = \sqrt{a^2 + b^2}$$

$$\beta = \arctan(b, a)$$

$$a = d \cos(\beta)$$

$$b = d \sin(\beta)$$



Example: Inverse Kinematics AIBO

$$x = l_2 \cos(\theta_1) \sin(\theta_3) + l_2 \sin(\theta_1) \cos(\theta_2) \cos(\theta_3) + l_1 \sin(\theta_1) \cos(\theta_2)$$

$$= a \cos(\theta_1) + b \sin(\theta_1)$$

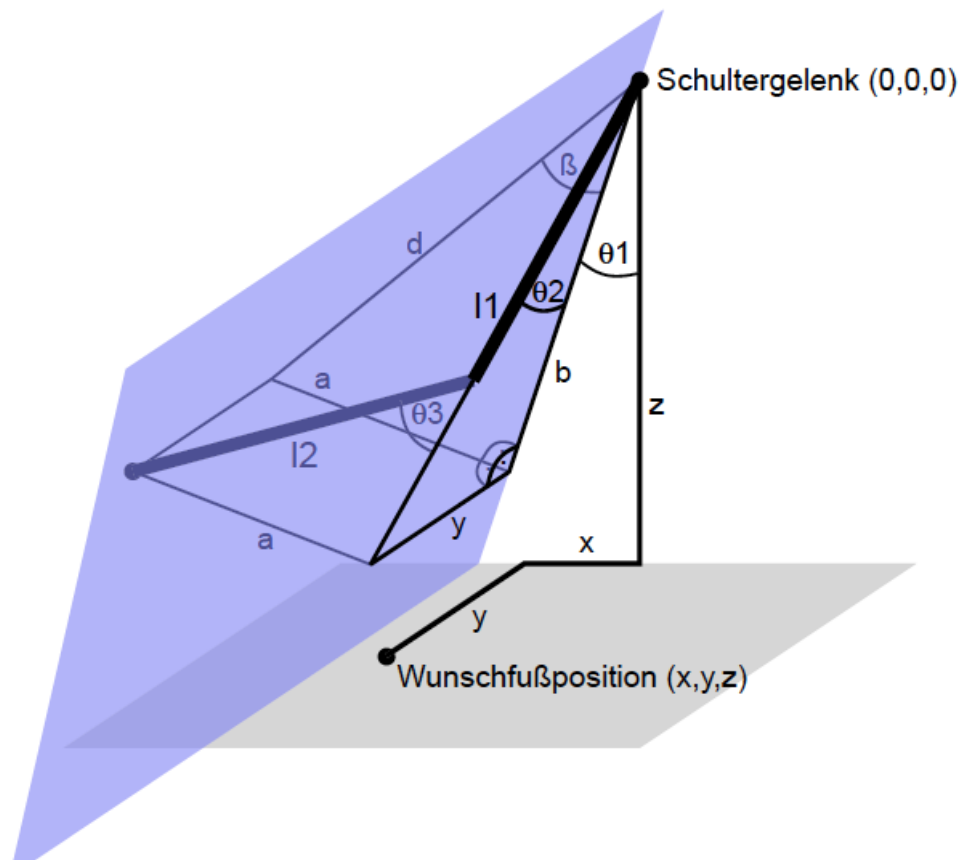
$$= d \cos(\theta_1) \cos(\beta) + d \sin(\theta_1) \sin(\beta)$$

$$= d \cos(\theta_1 + \beta)$$

$$z = d \sin(\theta_1 + \beta) \quad .$$

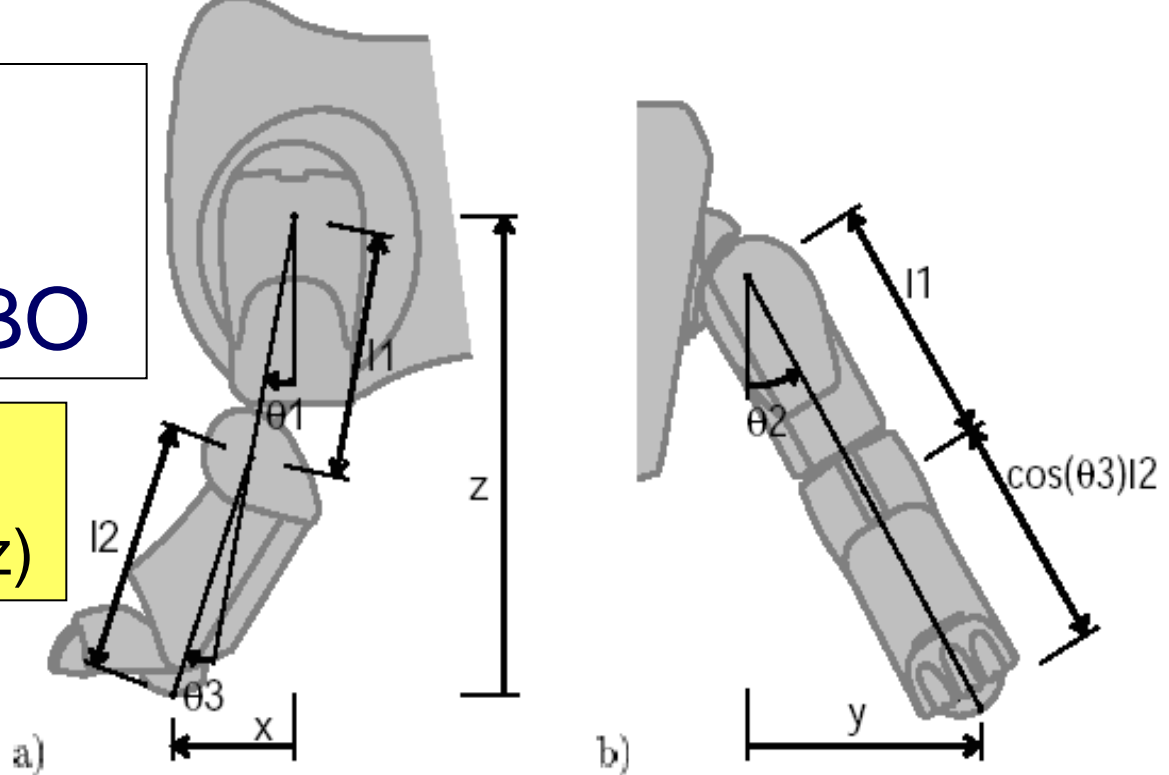
$$\theta_1 + \beta = \arctan(z, x)$$

$$\theta_1 = \arctan(z, x) - \beta$$



Example: Inverse Kinematics AIBO

Calculate q_1, q_2, q_3
for feet position (x, y, z)



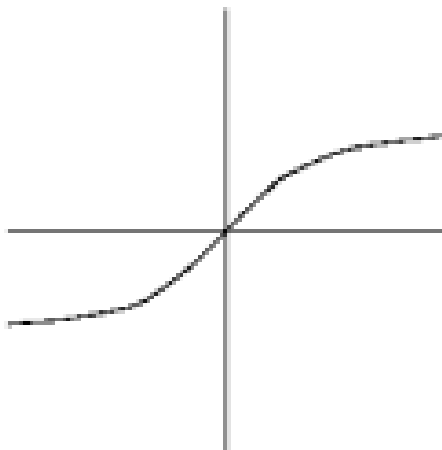
$$\theta_3 = \arccos \left(\frac{x^2 + y^2 + z^2 - l_1^2 - l_2^2}{2l_1l_2} \right)$$

$$\theta_2 = \arcsin \left(\frac{y}{l_1 + l_2 \cos(\theta_3)} \right)$$

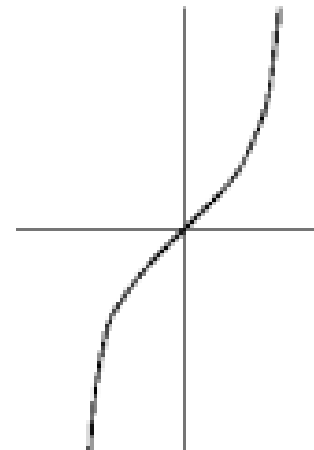
$$\theta_1 = \arctan(z, x) - \arctan \left((l_1 + l_2 \cos(\theta_3)) \cos(\theta_2), l_2 \sin(\theta_3) \right)$$

Special Benefits in Calculations

- Rotations in a plane (around joint axis)
- Select "simple" solutions
- Select "simple" relationships
- Use arctan (better: atan2) instead of arcsin or arccos (because of large error propagation near -1/+1)



arctan



arcsin

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Kinematics of Drive Systems

Kinematics (forward kinematics):

- Where does it move to?

Inverse kinematics (reverse kinematics):

- How can it get there?



Simplification:
Neglect mass and force

Kinematics of Drive Systems

- Driven wheels or chains
- Further wheels as stabilizers or for odometry
- Controlable wheels

Idealizing assumptions:

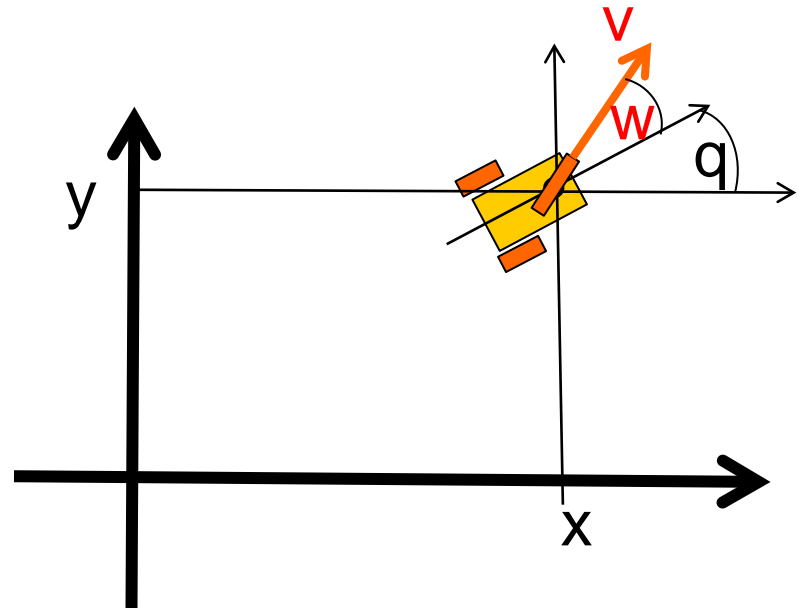
- Wheels run straight (perpendicular to the axis)
- Forward movement per complete rotation: $2\pi r$ for radius r
- Forward movement per rotation about w : wr for radius r

Drives for Vehicles on a Plane

Work space:

Pose (x, y, q) with 3 DOF

$q = 0$ in x-direction



$V(t) = (V_x(t), V_y(t))$ and $w(t)$ are *control parameters* for motion. They depend on position and speeds of driving wheels.

Kinematics/Inverse Kinematics

Kinematics: Calculate motion from control.

Change from pose $(0,0,0)$ to $(x(t),y(t),q(t))$
by speed $V(t) = (V_x(t), V_y(t))$ in direction $w(t)$

$$\begin{aligned}x(t) &= \int_0^t V_x(t) dt = \int_0^t V(t) \cos [q(t)] dt \\y(t) &= \int_0^t V_y(t) dt = \int_0^t V(t) \sin [q(t)] dt \\q(t) &= \int_0^t w(t) dt\end{aligned}$$

Inverse Kinematics:

Which control V and w is needed for desired motion?
Options depend on kind of drive.

Drives for Vehicles on a Plane

Configuration space:

Options for control:

- Speeds of the driving wheels
- Directions of the wheels / axes

Limitations by constraints

e.g.

- connections between wheels
- Dependency between direction and speed of wheels



ICC = instantaneous center of curvature

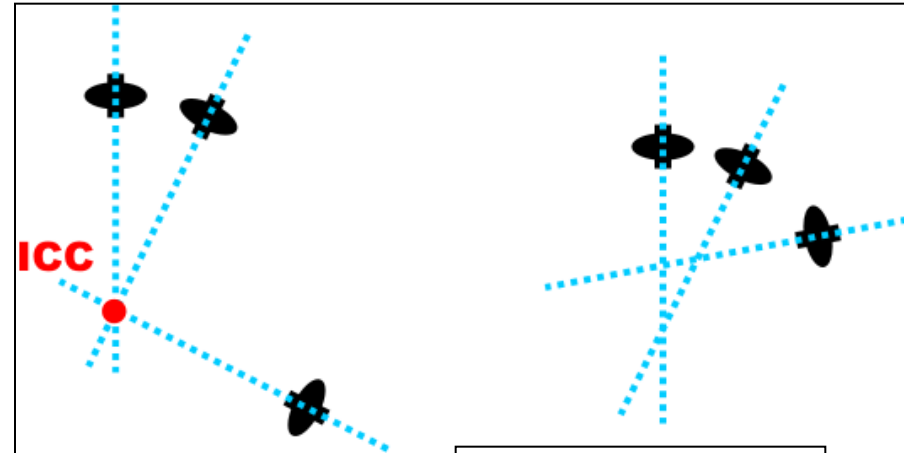
ICC defined as intersection point of all axes

Constraints for smooth motion:

- ICC exists
- Consistent speed of driving wheels

Otherwise:

- Robot loses traction
- Robot slides, unpredictable motion



Images from
Borenstein et.al.:
Where am I?

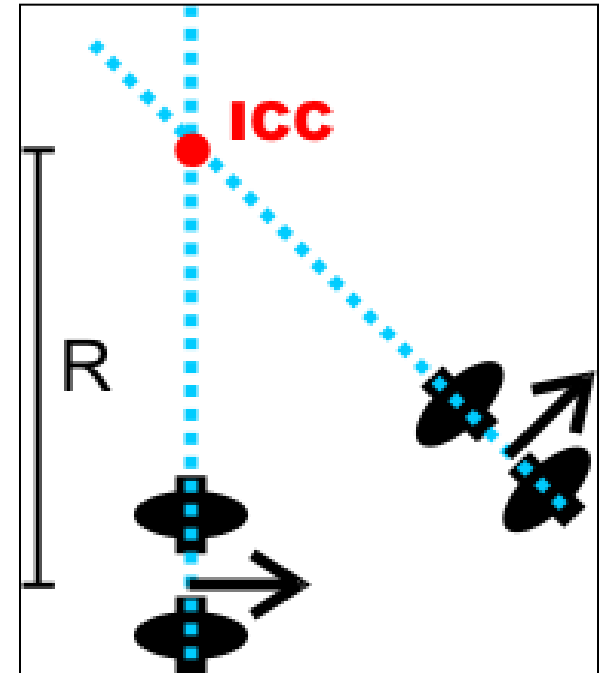
ICC = instantaneous center of curvature

Robot moves on a circle around ICC.

(Straight move for parallel axes:
ICC infinitely far.)

ICC can be changed by

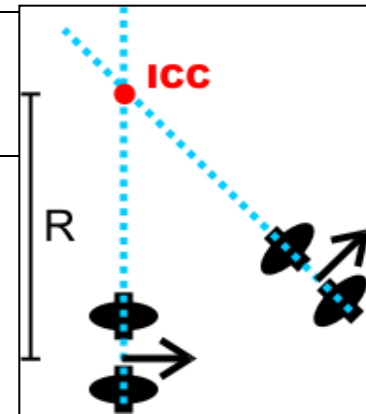
- steering of axes/wheels
- different speeds of driving wheels



Kinematics by ICC

Position of ICC for robot at pose (x, y, q) :

$$ICC = [x - R \sin(\theta), y + R \cos(\theta)]$$



Pose of Robot after time dt

while robot rotates ωdt around ICC:

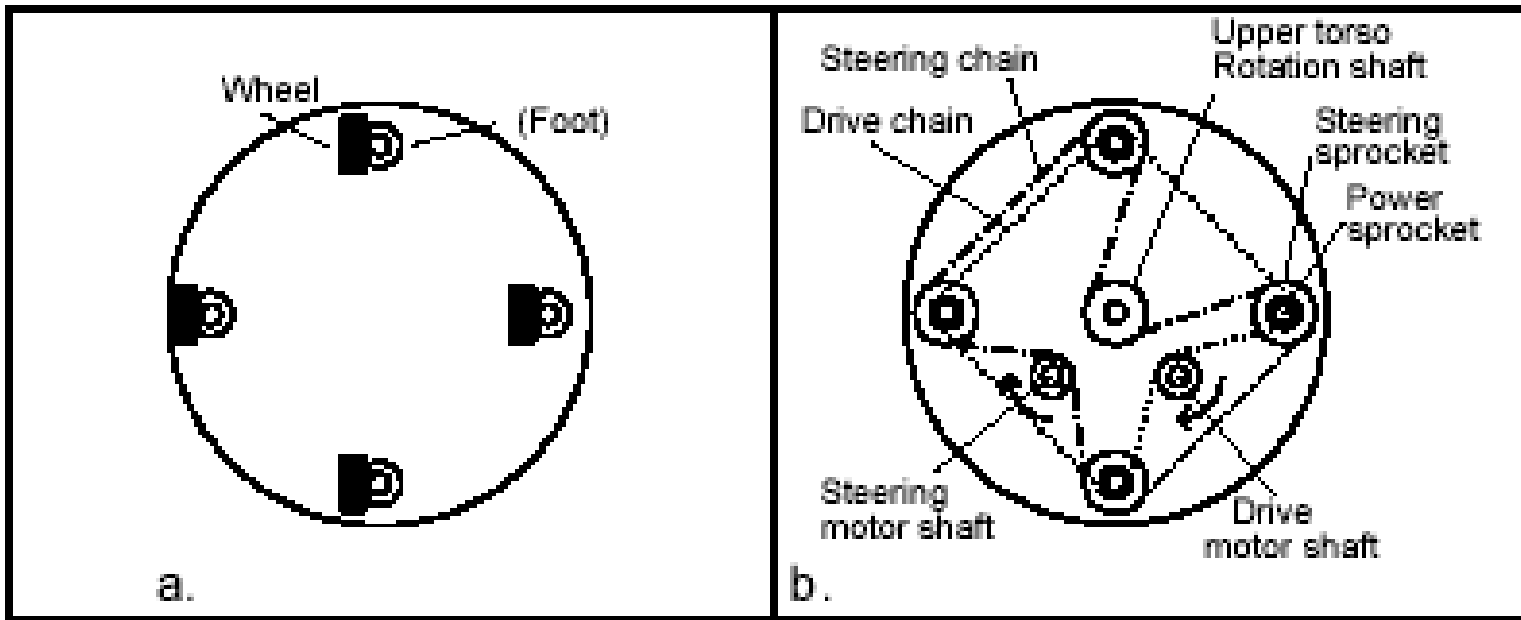
$$\begin{bmatrix} x' \\ y' \\ \theta' \end{bmatrix} = \begin{bmatrix} \cos(\omega \delta t) & -\sin(\omega \delta t) & 0 \\ \sin(\omega \delta t) & \cos(\omega \delta t) & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x - ICC_x \\ y - ICC_y \\ \theta \end{bmatrix} + \begin{bmatrix} ICC_x \\ ICC_y \\ \omega \delta t \end{bmatrix}$$

Synchrodrive

All wheels in same steerable direction w with identical speed.
ICC infinitely far perpendicular to direction w

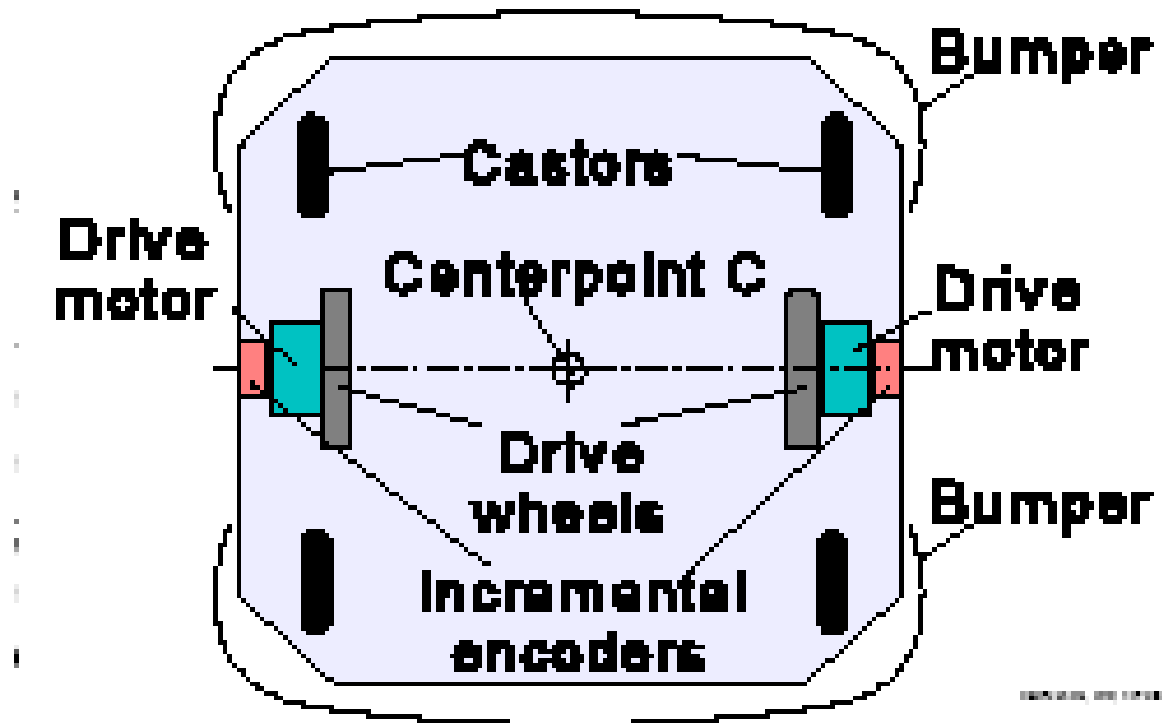
Control:

Speed v and direction w
of wheel(s)



Differential Drive

Driving wheels on 1 axis with different speeds



Differential Drive

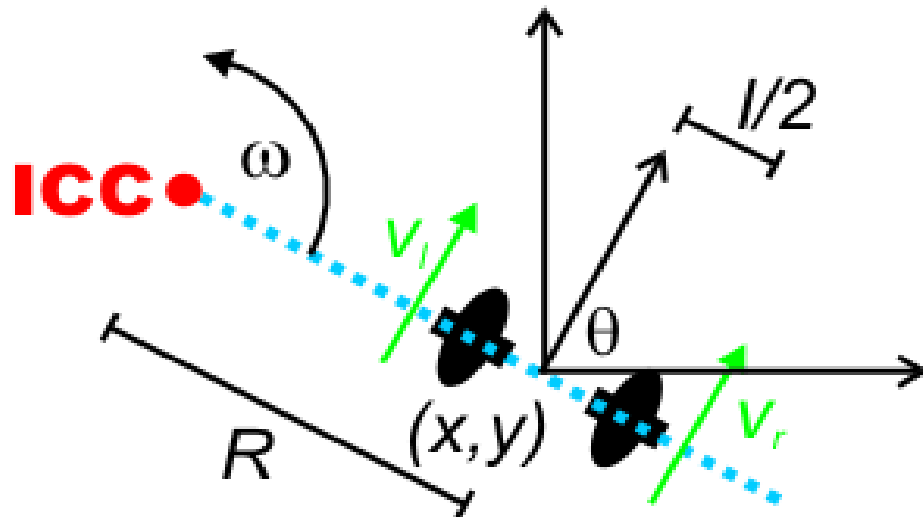
ICC on the axis, position depends on v_l, v_r

$v_l = v_r$ moves straight on

$v_l = -v_r$ turns around

Control:

Speeds v_l and v_r



$$\omega(R + l/2) = v_r$$

$$\omega(R - l/2) = v_l$$

$$R = \frac{l}{2} \frac{(v_l + v_r)}{(v_r - v_l)}$$

$$\omega = \frac{v_r - v_l}{l}$$

Differential Drive: Kinematics

Change from pose $(0,0,0)$ to $(x(t),y(t),q(t))$
by speeds v_l and v_r of left and right wheel

$$x(t) = \frac{1}{2} \int_0^t [v_r(t) + v_l(t)] \cos[\theta(t)] dt$$

$$y(t) = \frac{1}{2} \int_0^t [v_r(t) + v_l(t)] \sin[\theta(t)] dt$$

$$\Theta(t) = \frac{1}{l} \int_0^t [v_r(t) - v_l(t)] dt$$

Differential Drive: Inverse Kinematics

Which controls $v_l(t), v_r(t)$ result in undesired motion?

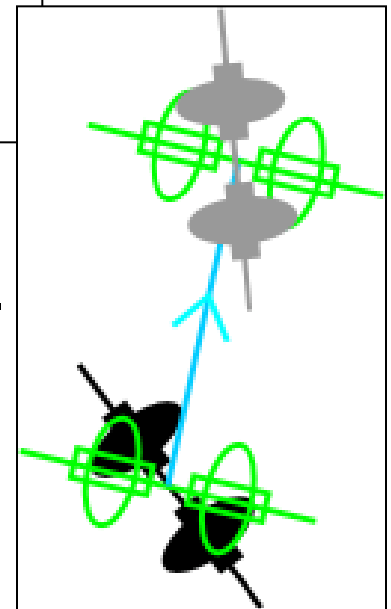
$$x(t) = \frac{1}{2} \int_0^t [v_r(t) + v_l(t)] \cos[\theta(t)] dt$$

$$y(t) = \frac{1}{2} \int_0^t [v_r(t) + v_l(t)] \sin[\theta(t)] dt$$

$$\Theta(t) = \frac{1}{l} \int_0^t [v_r(t) - v_l(t)] dt$$

Many different solutions to arrive at a given target.

No motion in direction of the axis (towards ICC).



Differential Drive: Inverse Kinematics

$$x(t) = \frac{1}{2} \int_0^t [v_r(t) + v_l(t)] \cos[\theta(t)] dt$$

$$y(t) = \frac{1}{2} \int_0^t [v_r(t) + v_l(t)] \sin[\theta(t)] dt$$

$$\Theta(t) = \frac{1}{l} \int_0^t [v_r(t) - v_l(t)] dt$$

Special cases:

$$v = v_l = -v_r :$$

$$\begin{pmatrix} x' \\ y' \\ \theta' \end{pmatrix} = \begin{pmatrix} x \\ y \\ \theta + 2v\delta t/l \end{pmatrix}$$

Turn on place

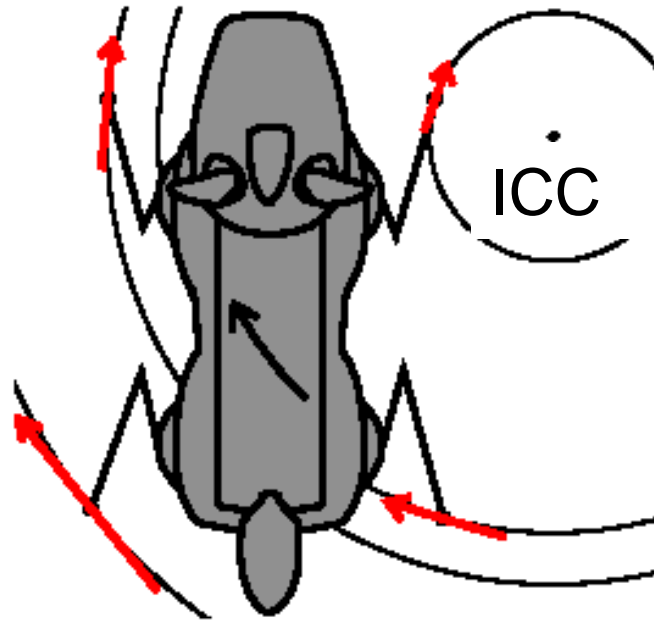
$$v = v_l = v_r :$$

$$\begin{pmatrix} x' \\ y' \\ \theta' \end{pmatrix} = \begin{pmatrix} x + v \cos(\theta)\delta t \\ y + v \sin(\theta)\delta t \\ \theta \end{pmatrix}$$

Forward motion

AIBO: „Wheel model“ (Differential Drive)

Curved motion by different speeds of legs.

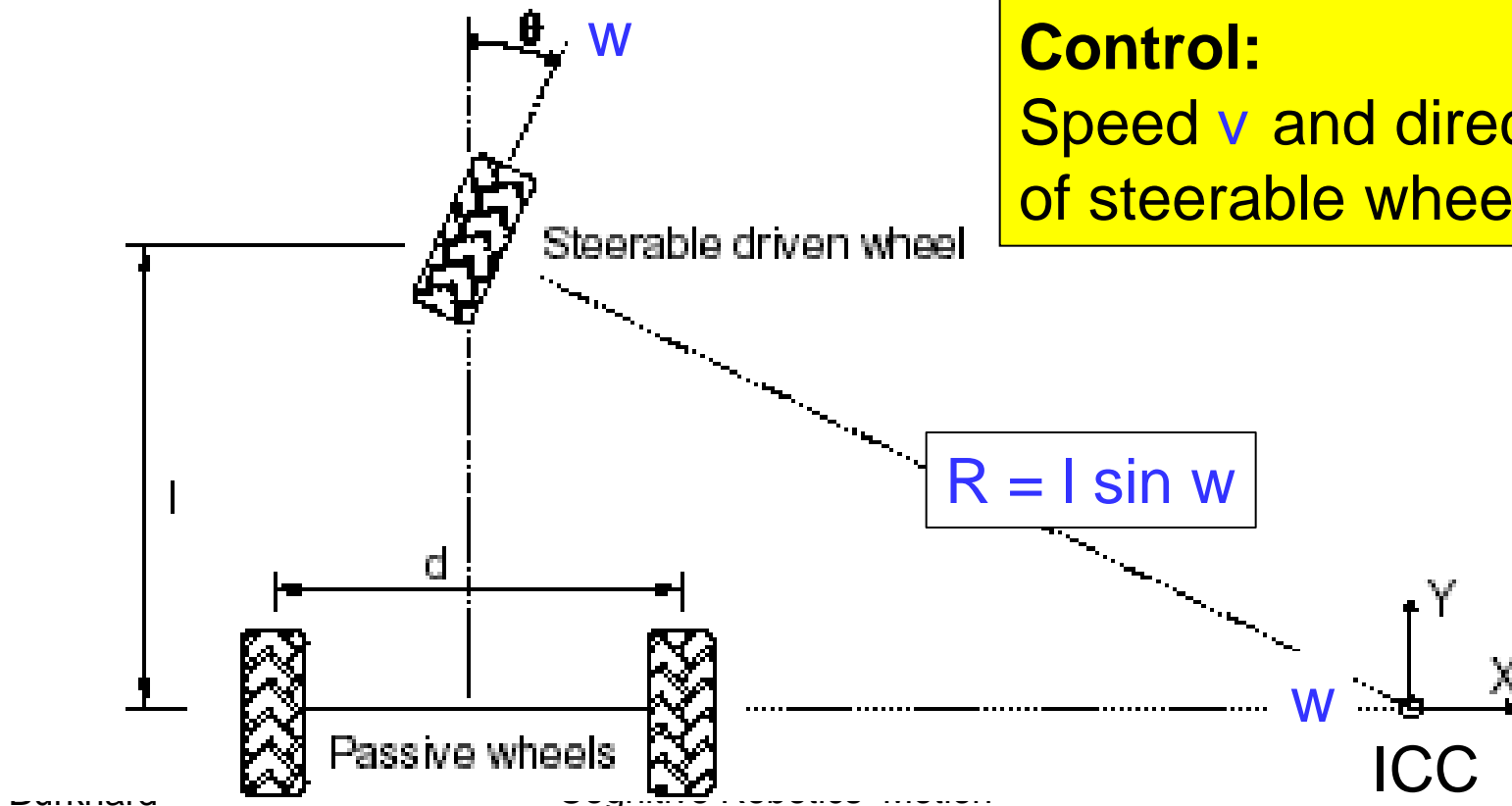


Controlled Wheels

One (or more connected) steerable wheels, other wheels passive:
Bicycle, Tricycle, Wagon etc.

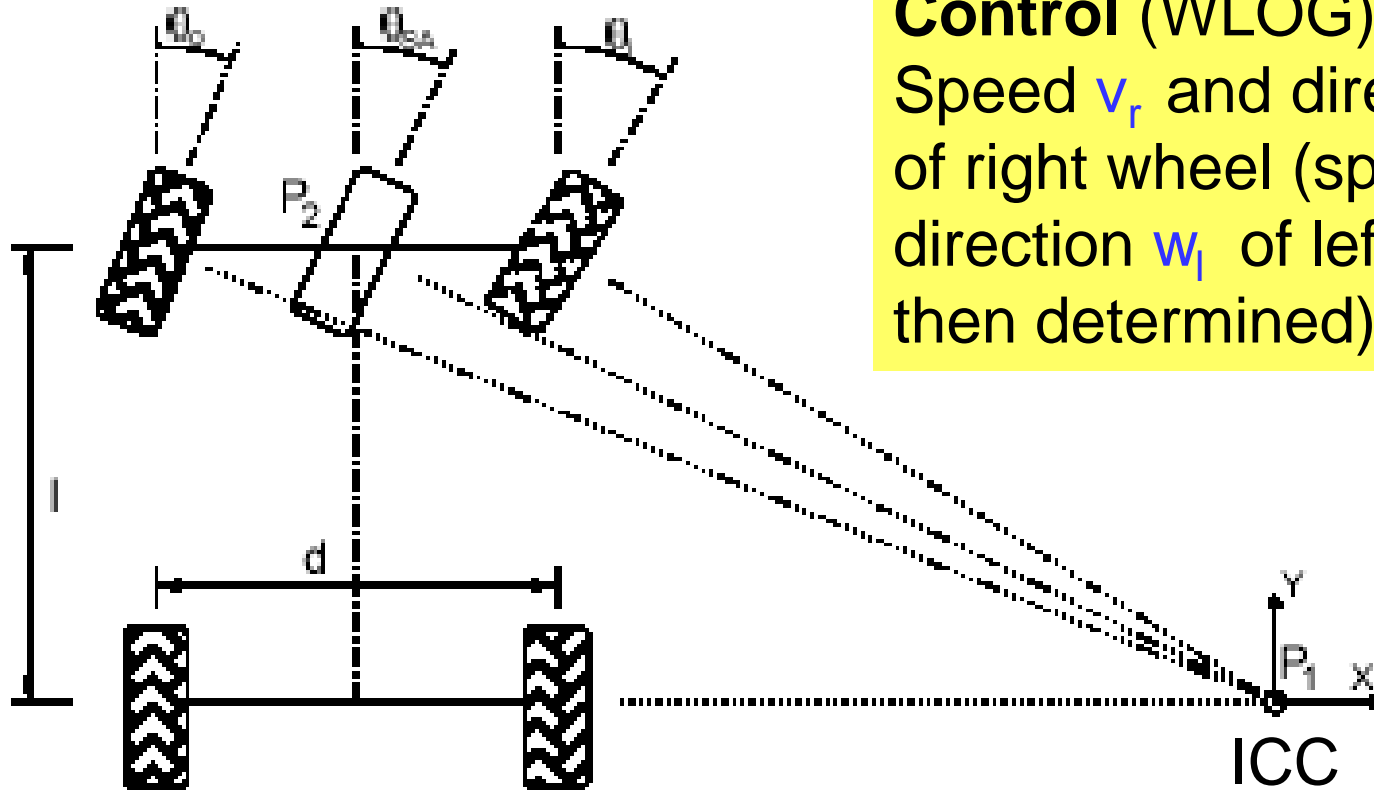
ICC on the axis of passive wheel(s), position depends on w

Control:
Speed v and direction w
of steerable wheel(s)



Ackermann-Drive: Automobile

Front wheels are individually steerable



Control (WLOG):
Speed v_r and direction w_r
of right wheel (speed v_l and
direction w_l of left wheel is
then determined)

Modell with ICC like for tricycle by a phantom wheel at P_2

Characteristica of Drives:

Mostly 2 control parameters
for 3 spatial DOF
Ø Nonholonomic drives

Rotation on place:

- Differential drive
- Tricycle, Ackerman only for $w = 90^\circ$ (with stability problems)

Differential drive:

- Uneven terrain and sliding results in direction errors for.

Tricycle, Ackerman:

- Complicated maneuvers (parking!)

Ackerman:

- Improved stability by separated (and slanted) front wheels

Degrees of Freedom (DOF) - continued

DOF is in both work space resp. configuration space the

- minimal number of parameters for complete description
- equivalently:
- maximal number of independent parameters

Work space: *effective DOF*

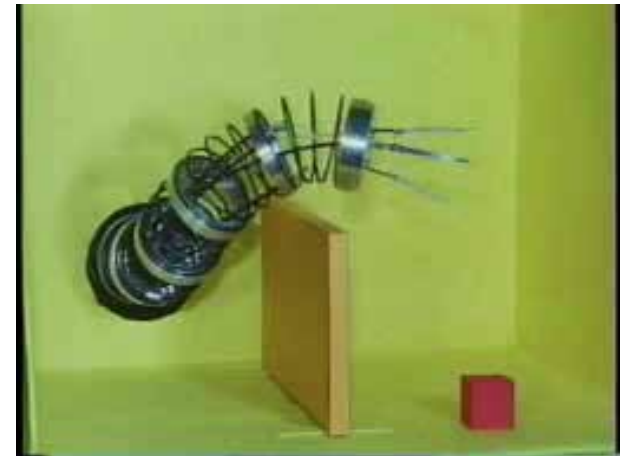
Configuration space: *controlable DOF*

Degrees of Freedom (DOF) - continued

Number of effective DOF

i.g. different from number of controllable DOF.

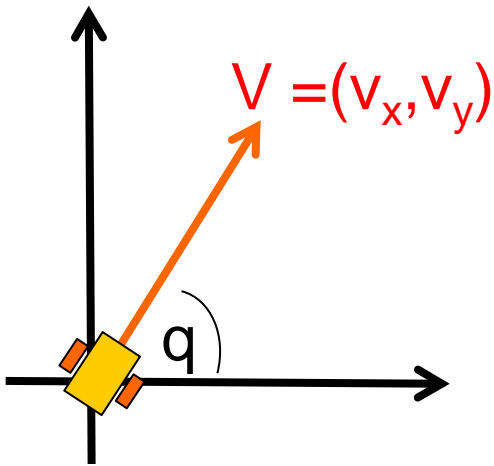
- All poses in work space may be reachable even in case of effective DOF $>$ controllable DOF (e.g. differential drive)
- effective DOF $<$ controllable DOF is useful in case of obstacles



Nonholonomic Drive Systems

Nonholonomic Constraints $C(p_1, \dots, p_n, \dot{p}_1, \dots, \dot{p}_n, t) = 0$
impose dependencies of parameters *and their derivatives*.

Holonomic Constraints
 $C(p_1, \dots, p_n, t) = 0$
impose dependencies of parameters.



Nonholonomic Constraint:

$$\tan q = v_y / v_x \quad \text{i.e.} \quad v_x \sin q - v_y \cos q = 0$$

$\cos q = 0$ for $q = \pi/2$ implies $v_x = 0$:

No motion in direction of axis (e.g. for differential drive)

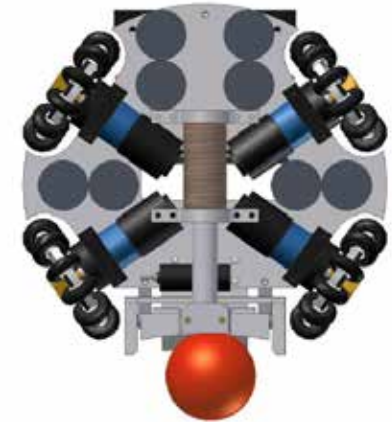
Holonomic Drive Systems

Most drive systems are nonholonomic and have only 2 controllable parameters

Fufighters
FU Berlin

Holonomic drives:

- Omnidirectional drive
(Control by separate motors of wheels)
- Synchrodive for rotationally symmetric vehicles
(2 spatial DOF)
- Synchrodive with additional body rotation



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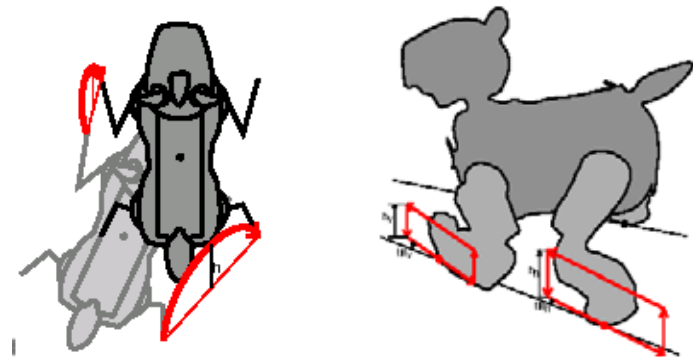
Trajectories

Trajectory in work space/configuration space:

Sequence of spatial parameters (positions/poses of the robot or its parts) or of control parameters at different times,

e.g.

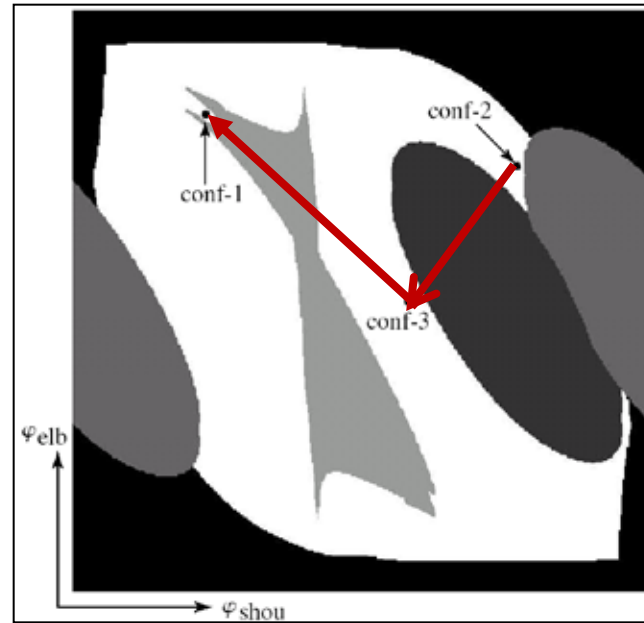
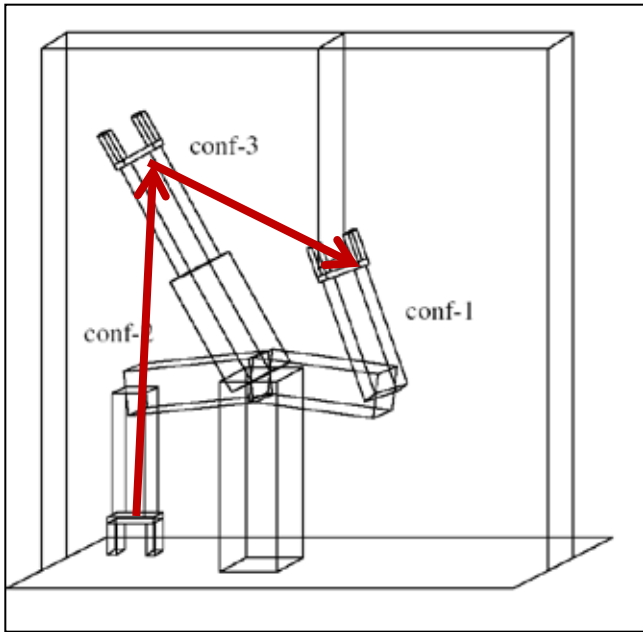
- trajectory of CoM (center of mass)
- trajectory of feet
- trajectory of limb angles



Diploma Thesis U. Duffert

Trajectories

Set of poses $p(t)$ and corresponding configurations $q(t)$:



Motion planning: Find realistic (and optimal) trajectories.
The trajectories in the figures are not realistic.

Trajectories of Keyframes

Sequence of Keyframes:

Characteristic poses during a motion (“like in a comic”).

Originally used in animated movies.

Transition times define speed to reach next pose.

Poses between keyframes must be interpolated.

Keyframe

Time 1000

HeadPitch HeadYaw 0

RShoulderPitch LShoulderPitch 120

RShoulder RollLShoulderRoll 0

RElbowRoll 90

LElbowRoll -90

RElbowYaw 90

LElbowYaw -90

RHipYawPitch LHipYawPitch 0

RHipPitch LHipPitch -31

RHipRoll LHipRoll 0

RKneePitch LKneePitch 63

RAnklePitch LAnklePitch -31

....

Complete set of joint angles
to be set in given time

Motion Skill: Sequence of Keyframes

```
300 0 -21 -62 32 -69 -59 0 -8 12 -10 -0 12 -11 0 8 12 -0 -3 -11 -110 -32 69 59
300 -5 -21 -62 46 -69 -59 0 0 18 -0 -9 -4 0 0 -10 -0 17 -5 -62 -46 69 59
300 0 -21 -62 60 -69 -59 0 8 -10 -0 12 -11 0 8 12 -0 -3 -11 -110 -32 69 59
300 0 -21 -75 60 -69 -59 0 8 6 -36 27 -11 0 8 12 -15 7 -11 -97 -32 69 59
300 0 -21 -86 60 -69 -59 0 8 42 -69 13 -11 0 8 12 -30 23 -11 -86 -32 69 59
300 0 -21 -110 60 -69 -59 0 8 12 -0 -9 -11 0 8 -10 -0 12 -14 -62 -32 69 59
300 -5 -21 -110 46 -69 -59 0 0 18 -0 -9 -4 0 0 -10 -0 17 -5 -62 -46 69 59
300 0 -21 -110 32 -69 -59 0 -8 12 -0 -3 11 0 -8 -10 -0 12 11 -62 -60 69 59
300 0 -21 -97 32 -69 -59 0 -8 12 -15 7 11 0 -8 6 -36 27 11 -75 -60 69 59
300 0 -21 -84 32 -69 -59 0 -8 12 -30 23 11 0 -8 42 -69 13 11 -84 -60 69 59
```

FILE walk_forward-flemming-nika.txt
in ../keyframes

Each line starts with the transition time followed by the target angles of joints in a predefined order.

RoboNewbie:

Keyframe sequences are “played” by class `keyframeMotion`.

Order of Joints in RoboNewbie Keyframes

NeckYaw = 0

NeckPitch = 1

LeftShoulderPitch = 2

LeftShoulderYaw = 3

LeftArmRoll = 4

LeftArmYaw = 5

LeftHipYawPitch = 6

LeftHipRoll = 7

LeftHipPitch = 8

LeftKneePitch = 9

LeftFootPitch = 10

LeftFootRoll = 11

RightHipYawPitch = 12

RightHipRoll = 13

RightHipPitch = 14

RightKneePitch = 15

RightFootPitch = 16

RightFootRoll = 17

RightShoulderPitch = 18

RightShoulderYaw = 19

RightArmRoll = 20

RightArmYaw = 21

Keyframes

Simple implementation

Simple design (especially with “teaching”)

But motions can not adapt

Best suited for short sequences (stand-up, kick)

Usage of Trajectories for Motion Planning

Find a trajectory (path) of the robot or a part of the robot in work space or configuration space which satisfies certain conditions, e.g.

- Motion from start to destination while avoiding obstacles
- Motion of a limb while maintaining stability
- Motion of a manipulator to grasp an object

Side conditions may be time, energy, smoothness, stability, safety...

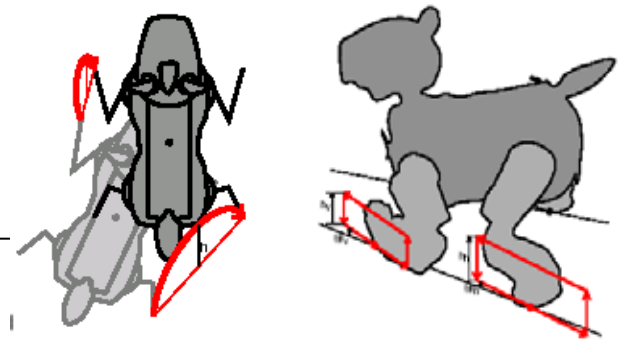
Appropriate trajectories can be found e.g. by physical models or by Machine Learning

Usage of Trajectories for Motion Control

Control the actuators (joint, limbs,...) such that the robot or a part of the robot follows a given trajectory.

Inverse kinematics

can be used to find the appropriate control parameters.



Shift CoM following a (straight) trajectory implies trajectories of feet, e.g. semi-ellipses or parallelograms. Related joint controls by inverse kinematics.

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Planning

... is a broad field in AI with many different methods.

Planning can be used for motions and for more complex behaviors (different time horizons).

Here: Some useful methods for motion planning

Later: Behavior planning

Motion Planning vs. Control

Robot can

- plan motions (and more complex behavior) before execution
- execution is then performed by appropriate control

Robot Control can be performed as

- Open loop control: ("blind" control)
Preplanned motions performed without sensor feedback.
- Closed loop control:
Sensor feedback is used for adaptation of intended motions.

Some planning methods lead directly to controls (e.g. potential fields).

Motion Planning vs. Control

Alternative for Planning:

Online motion control

by immediate reactions to sensor measurements
(e.g. for maintaining balance)

- sensor actor coupling
- behavioral robotics
- emergence principle

∅ later more

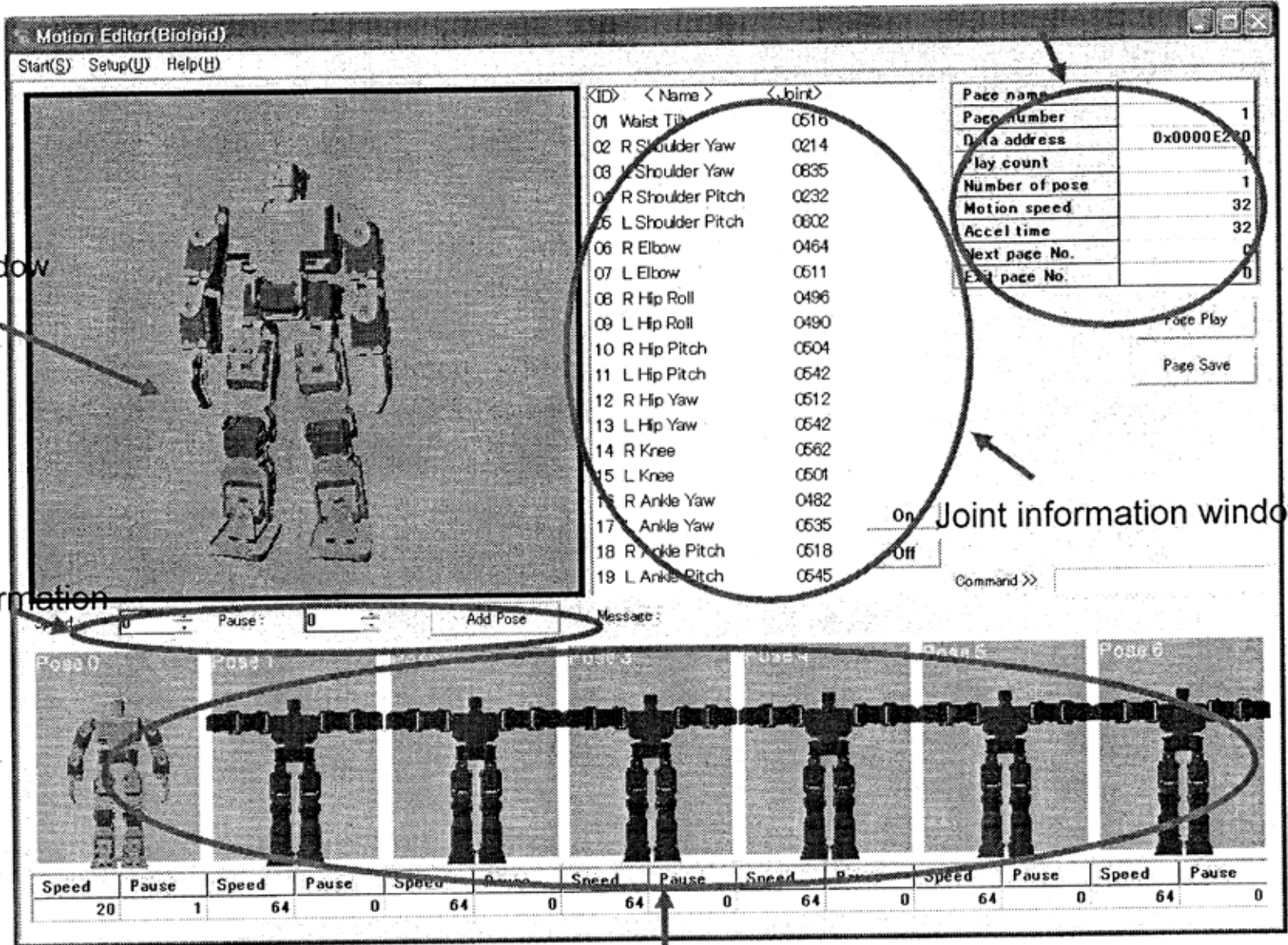
Teaching

- Set „characteristic poses“ of a motion by hand (at real robot) or by motion editor
- Protocol joint angles of each such pose as *keyframe* resulting in a sequence of keyframes
- Optimize (transition times, smoothing, ...) e.g. by machine learning



Motion Editor from Bioid Manual (2006)

Page information

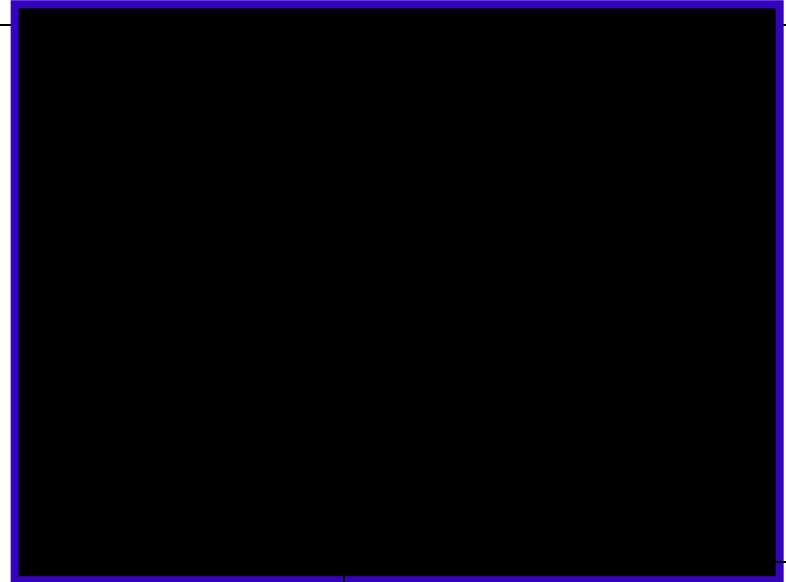


Joint information window

Saved pose window

Motion-Capturing

Imitate demonstrated motions



From IROS 2007

Markers at important points

Record motions (3D Motion Tracker)

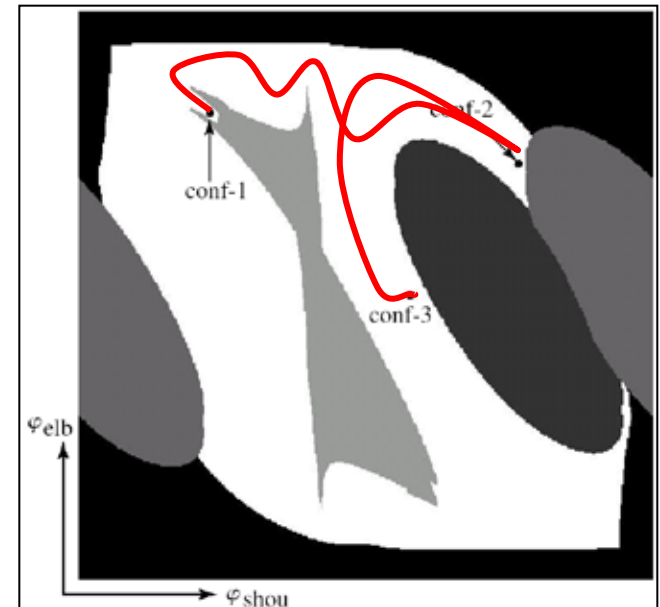
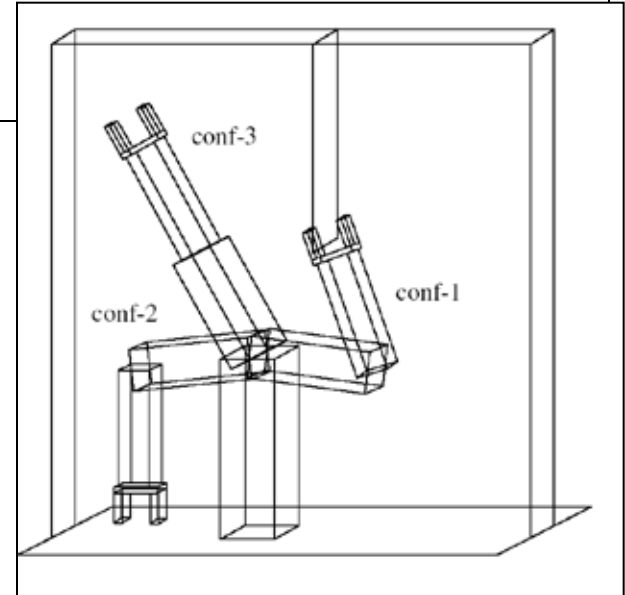
Implement related control (e.g. by analyzing the motion)

Motion Planning

Optimality of a trajectory may concern

- Length of path
- Time
- Smoothness
- Stability
- Safety
- Energy consumption
- Esthetics
- ...

Planning can be performed in work space or configuration space using path planning algorithms (e.g. A*)

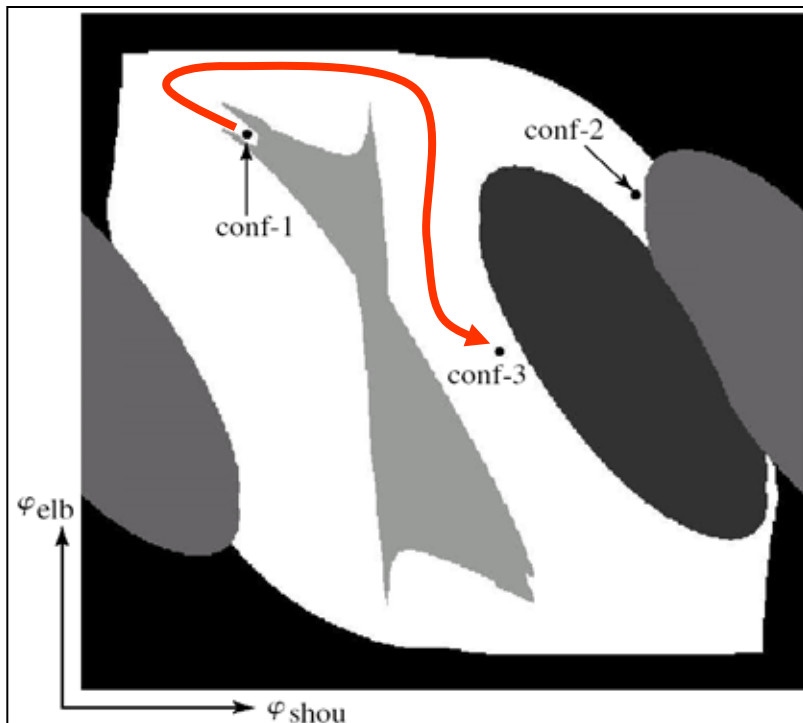


Images

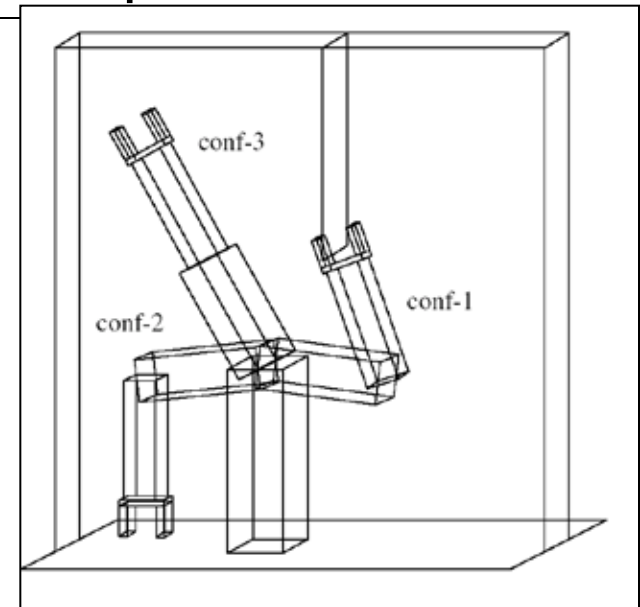
Planning in Configuration Space

Special regions in the configuration space for

- obstacles in work space (gray),
- geometry of the robot (black)

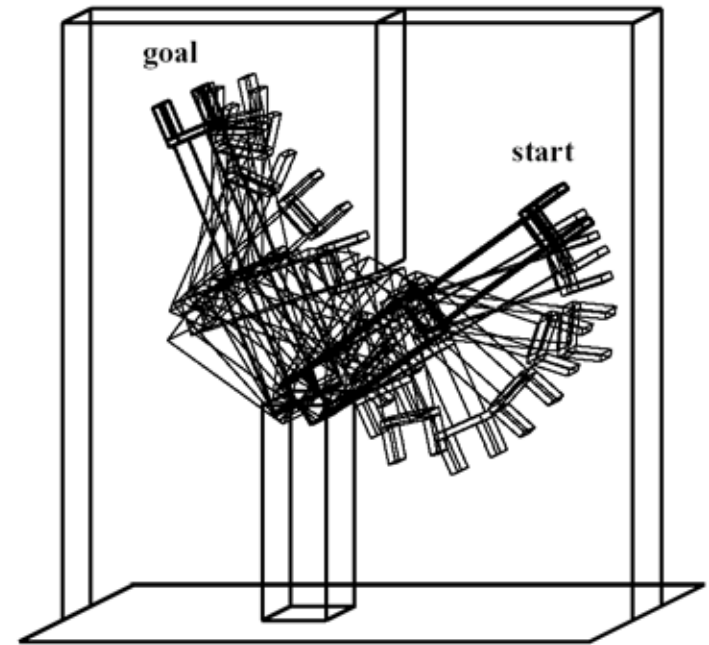
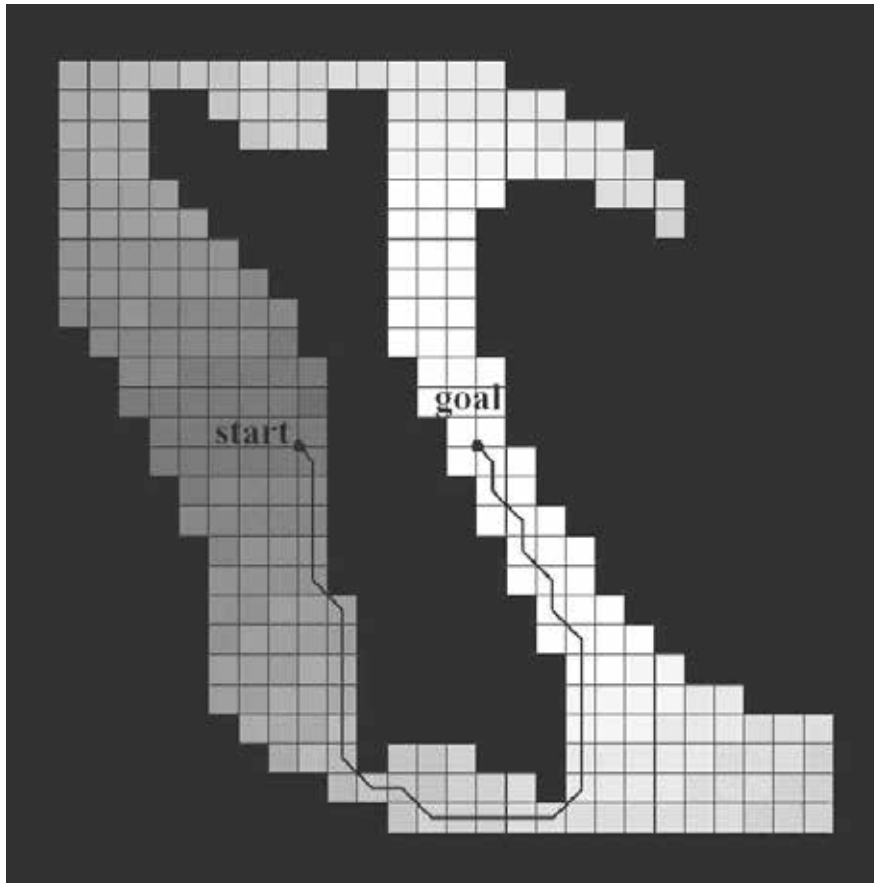


Results of planning in configuration space can be directly used as control for motion in work space.



Grid Based Search in Configuration Space

e.g. using graph search methods like A*

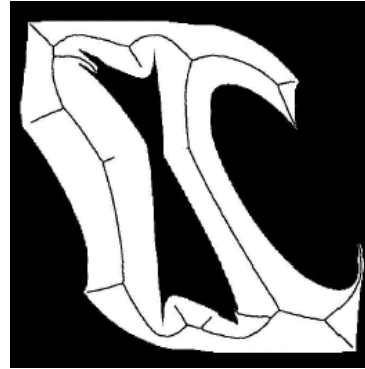


Skeleton Based Search in Configuration Space

Skeleton: Connects certain points.

Search for path on skeleton.

- as Voronoi-Graph:
points with equal minimal
distances to obstacles



- as Visibility Graph:
Nodes at corners of obstacles
Arcs between mutually observable nodes

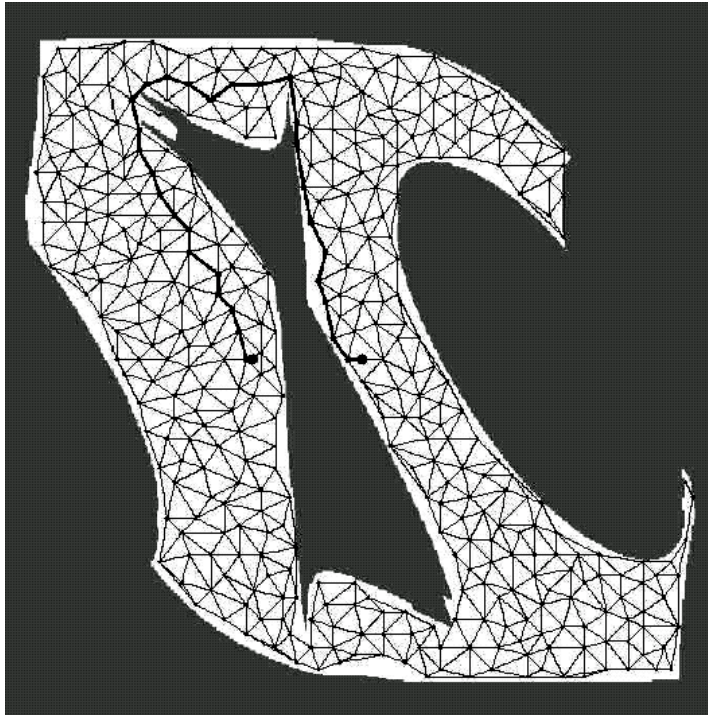
Problems:

- complex algorithms
- results often in detours

Random Point Search in Configuration Space

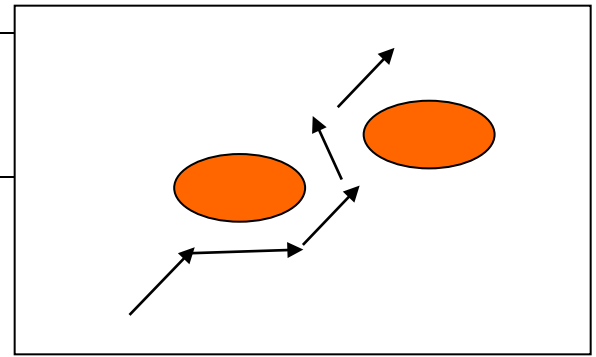
Graph search through random points in free space.

Ranking of preferable areas by differently distributed points.



Potential field

Control in a point (x_0, y_0)
as direction vector $[\mathbf{F}_x(x_0, y_0), \mathbf{F}_y(x_0, y_0)]$
of vector field $\mathbf{F}(x, y) = [\mathbf{F}_x(x, y), \mathbf{F}_y(x, y)]$



Special case:

Vector field $\mathbf{F}(x, y)$ is gradient of a potential field $\mathbf{U}(x, y)$

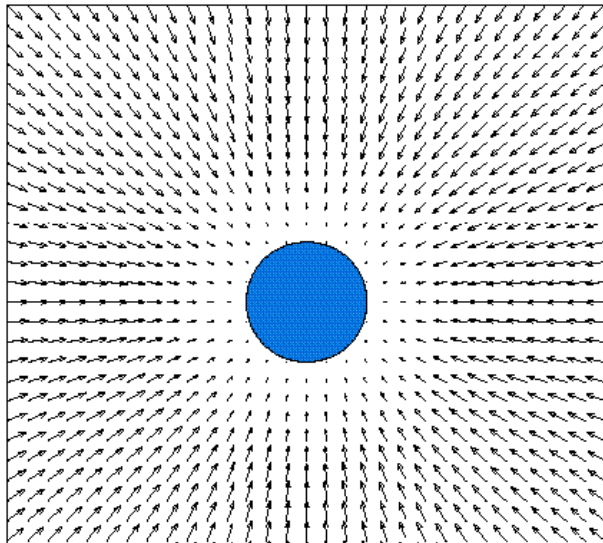
$$\mathbf{F}(x, y) = [d\mathbf{U}(x, y) / dx, d\mathbf{U}(x, y) / dy]$$

For application:

- Potential determined by environment/from sensory information
- Motion follows the gradient

Potential field

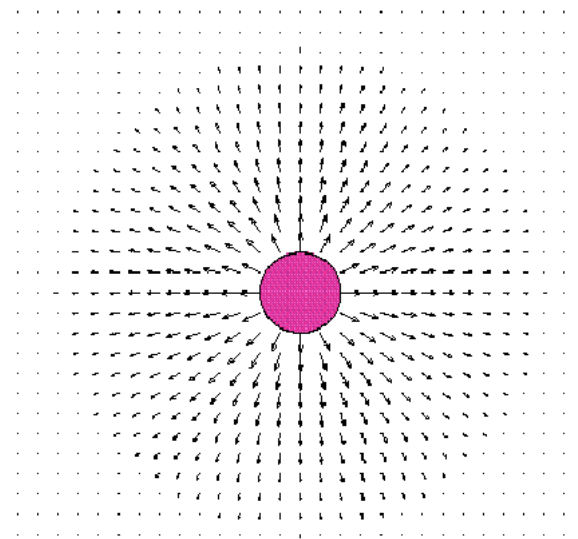
target:
attracting field



e.g.

$$U_{\text{goal}}(p) = a \text{ dist}(p, \text{goal})^2$$

obstacles:
repelling fields



e.g.

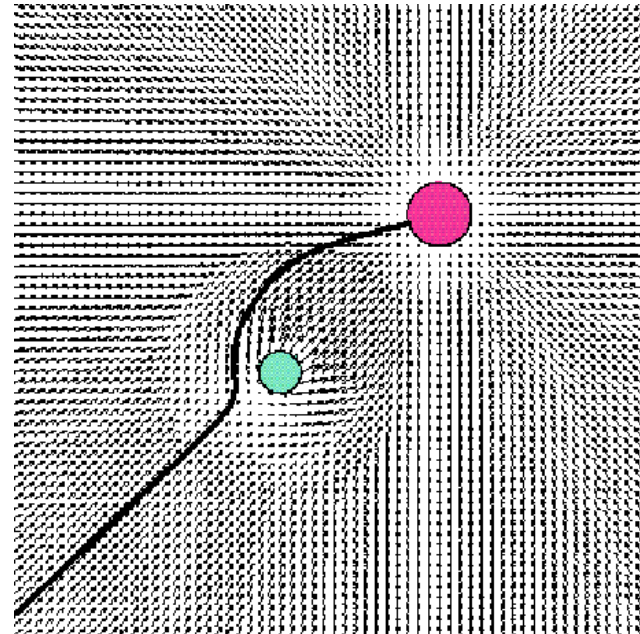
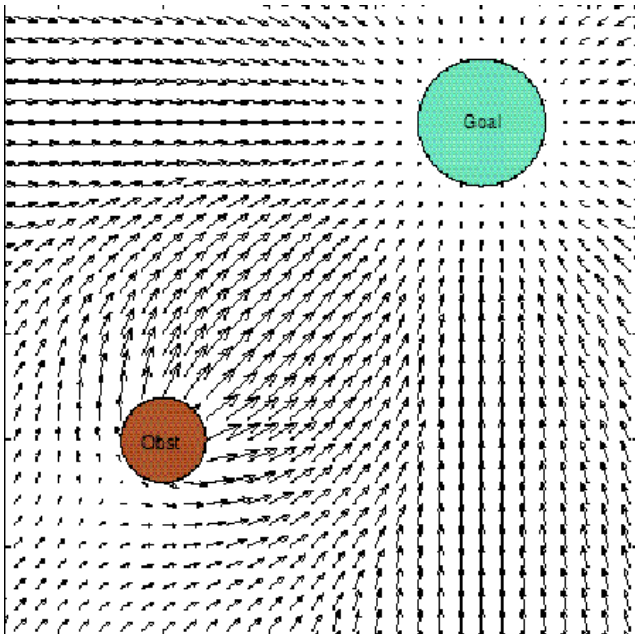
$$U_{\text{obstacle}}(p) = b \text{ dist}(p, \text{obstacle})^{-1}$$

Potential field

Potential field by superposition (addition):

$$\mathbf{U}(p) = \mathbf{U}_{\text{goal}}(p) + S \mathbf{U}_{\text{obstacle}}(p)$$

$$\mathbf{F} = - [d\mathbf{U} / dx , d\mathbf{U} / dy]$$



Potential field

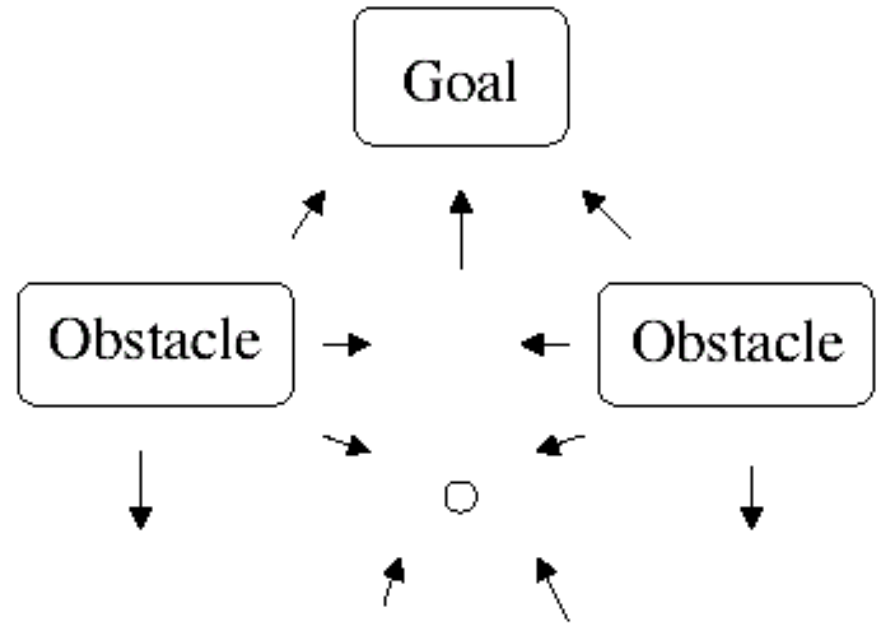
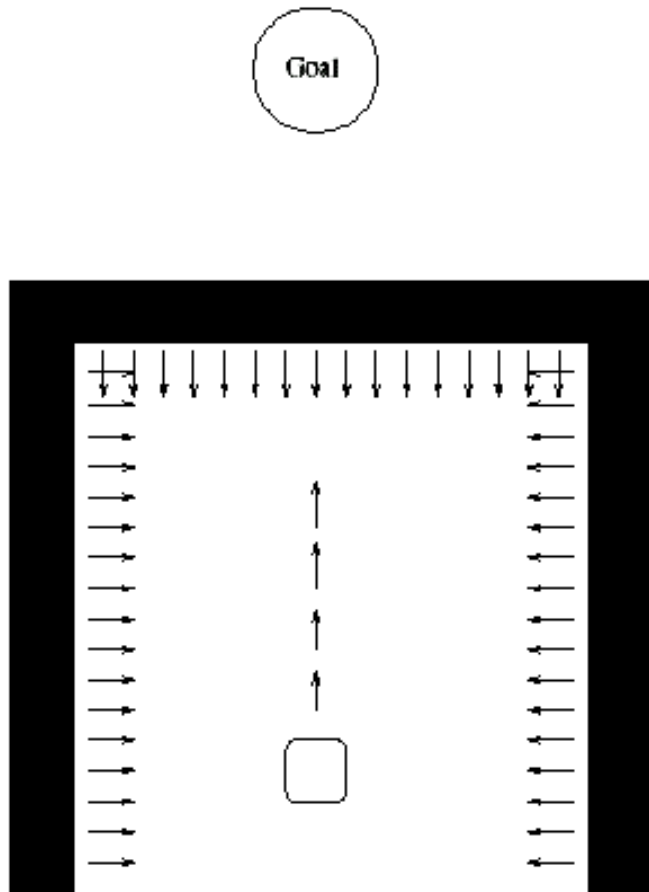
Benefits:

- direct usage for control
- local evaluation

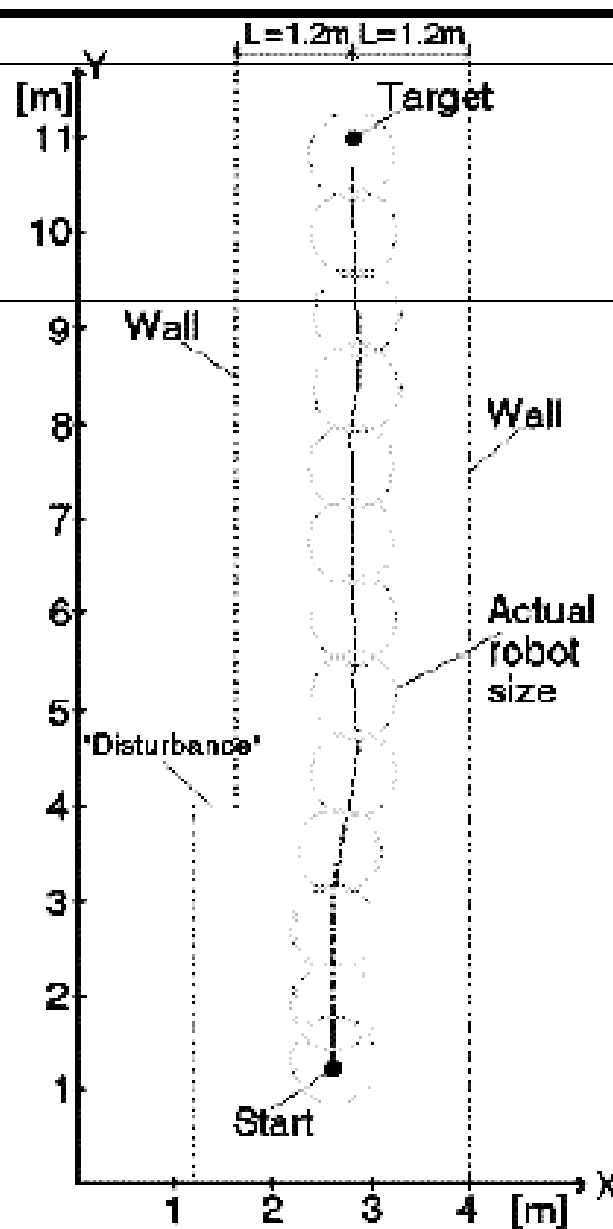
Problems:

- local minima
 - Compensation of fields,
 - "Trap" by close obstacles
- oscillating movements for
 - narrow areas
 - high speed

Potential field

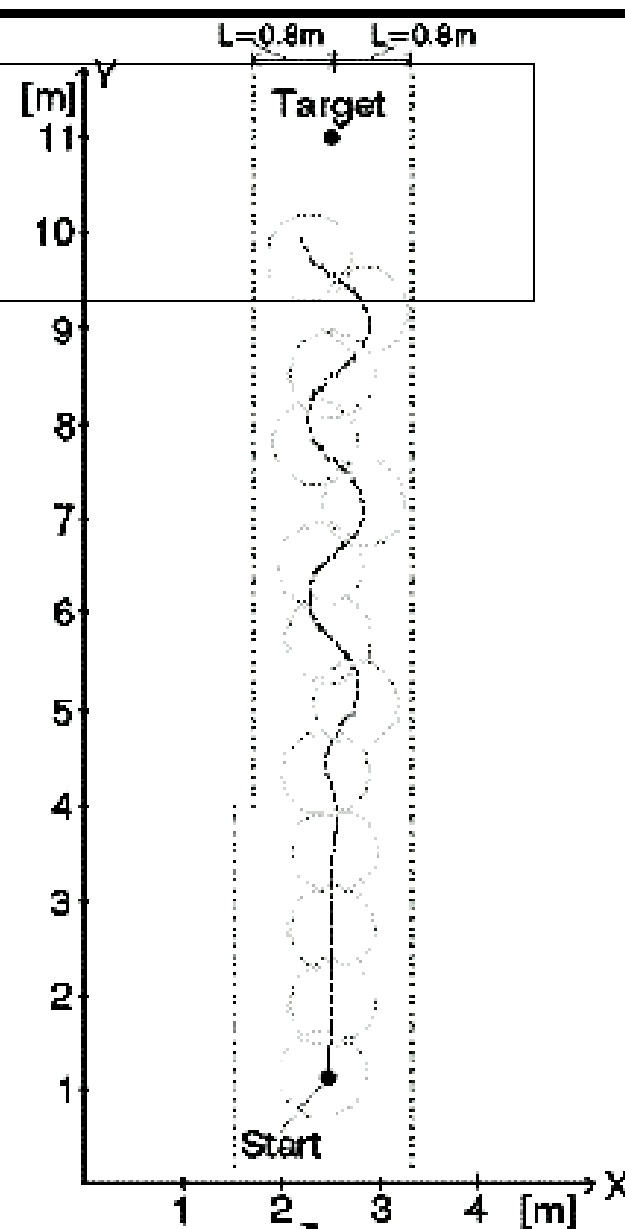


Potential field



a

Stable motion in wide corridor
 $V=0.8\text{m/s}$



b

Unstable motion in narrow
corridor. $V=0.8\text{m/sec.}$

Potential field

Additional other fields, e.g.

- rotating fields
- random fields

can

- specify directions
- break symmetries
- avoid (some) local minima

