

Dottorato di Ricerca in Ingegneria dei Sistemi

# **Control of Nonholonomic Systems**

Giuseppe Oriolo

*DIS, Università di Roma "La Sapienza"*

# LECTURE OUTLINE

## 1. Introduction

- nonholonomic systems? among the others
- kinematic constraints
- integrability of kinematic constraints
- a control viewpoint
- dynamics vs. kinematics
- more general nonholonomic constraints

## 2. Modeling Examples

- wheeled mobile robots
  - \* unicycle
  - \* car-like robot
  - \*  $N$ -trailer system
  - \* other wheeled mobile robots
- space robots with planar structure
  - \* two-body robot
  - \*  $N$ -body robot

### **3. Tools from Differential Geometry**

- Frobenius theorem
- integrability of Pfaffian constraints

### **4. Control Properties**

- controllability of nonholonomic systems
- stabilizability of nonholonomic systems
- classification of nonholonomic distributions
- examples of classification

### **5. Nonholonomic Motion Planning**

- chained forms
  - \* steering with sinusoidal inputs
  - \* steering with piecewise-constant inputs
  - \* steering with polynomial inputs
  - \* transformation into chained form
- WMRs in chained form
- unicycle simulation
- a general viewpoint: differential flatness

## 6. Feedback Control of Nonholonomic Systems

- basic problems
- asymptotic tracking
  - \* control properties
  - \* linear control design
  - \* nonlinear control design
  - \* dynamic feedback linearization
  - \* experiments with SuperMario
- posture stabilization: a bird's eye view

## 7. Optimal Trajectories for WMRs

(by M. Vendittelli)

- minimum time problems
- application to WMRs
  - \* extracting information from PMP
  - \* type A trajectories
  - \* type B trajectories

## REFERENCES

R. M. Murray, Z. Li, S. S. Sastry

*A Mathematical Introduction to Robotic Manipulation*, CRC Press, 1994

relevant for Lectures 1–5

A. De Luca, G. Oriolo\*

“Modelling and control of nonholonomic mechanical systems”, in *Kinematics and Dynamics of Multi-Body Systems* (J. Angeles, A. Kecskemethy Eds.), Springer-Verlag, 1995

relevant for Lectures 1–5

A. De Luca, G. Oriolo, C. Samson\*

“Feedback control of a nonholonomic car-like robot”, in *Robot Motion Planning and Control* (J.-P. Laumond Ed.), Springer-Verlag, 1998

relevant for Lectures 4–6

G. Oriolo, A. De Luca, M. Vendittelli

“WMR control via dynamic feedback linearization: Design, implementation and experimental validation”\*, *IEEE Transactions on Control System Technology*, vol. 10, no. 6, pp. 835–852, 2002.

relevant for Lectures 5–6

P. Souères, J.-D. Boissonnat†

“Optimal Trajectories for Nonholonomic Mobile Robots”, in *Robot Motion Planning and Control* (J.-P. Laumond Ed.), Springer-Verlag, 1998

relevant for Lecture 7

... and the references therein

\* downloadable from

<http://www.dis.uniroma1.it/~labrob/people/oriolo/oriolo.html>

† downloadable from

<http://http://www.laas.fr/~jpl/book-toc.html>

# INTRODUCTION

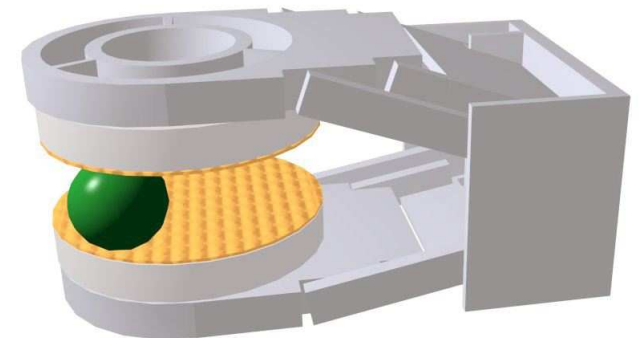
Nonholonomic systems? Among the others. . .



wheeled mobile robots (WMRs)

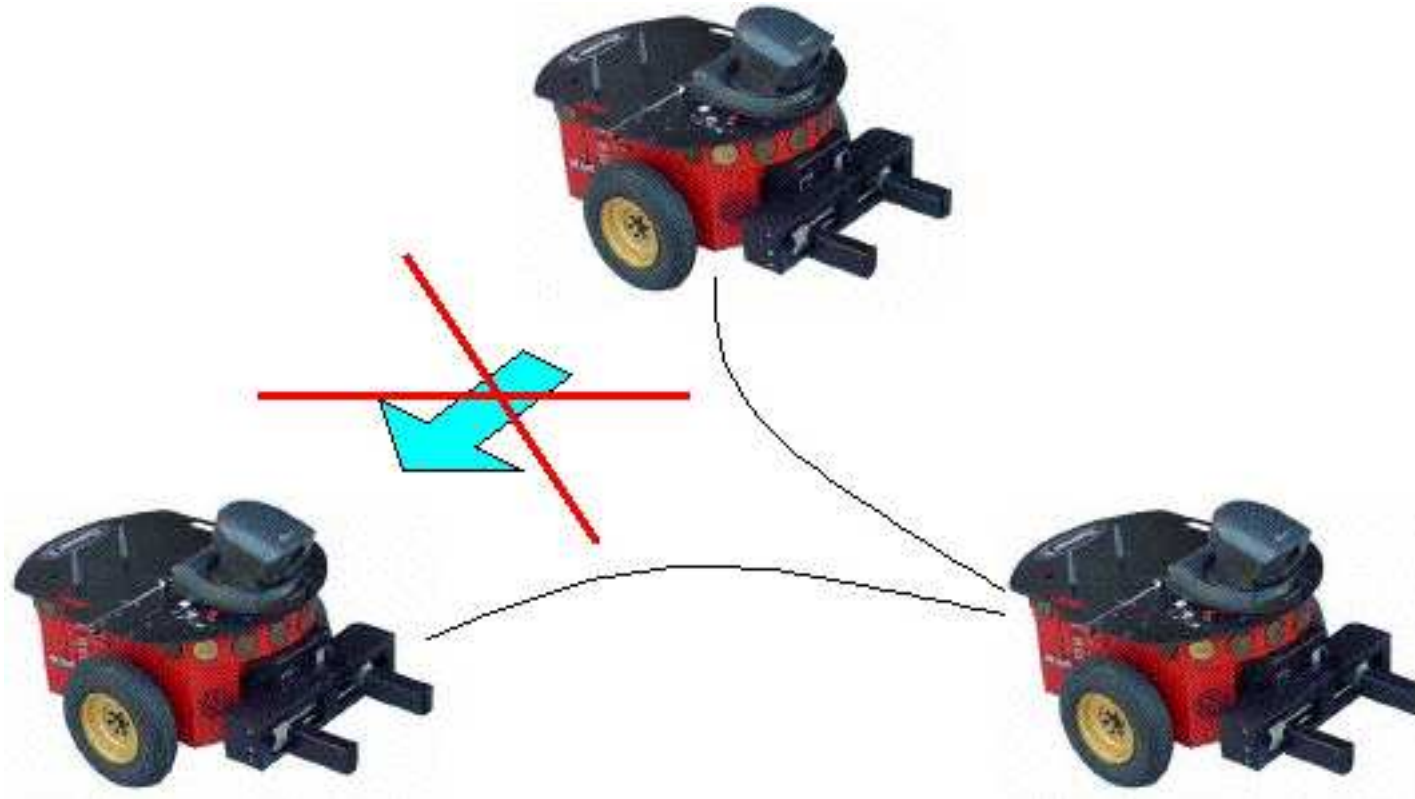


rolling manipulation



## what is nonholonomy?

due to the presence of wheels, a WMR **cannot move sideways**



this is the **rolling without slipping** constraint, a special case of **nonholonomic** behavior

in general: a **nonholonomic** mechanical system **cannot move in arbitrary directions** in its configuration space

problems:

- our everyday experience indicates that WMRs are controllable, but can it be proven?  
↔ we need a mathematical characterization of nonholonomy
- in any case, if the robot must move between two configurations, a **feasible** path is required (i.e., a motion that complies with the constraint)  
↔ we need appropriate path planning techniques
- the feedback control problem is much more complicated, because:
  - ◇ a WMR is **underactuated**: less control inputs than generalized coordinates
  - ◇ a WMR is **not smoothly stabilizable** at a point↔ we need appropriate feedback control techniques

## Kinematic Constraints

- the configuration of a mechanical system can be uniquely described by an  $n$ -dimensional vector of **generalized coordinates**

$$q = (q_1 \quad q_2 \quad \dots \quad q_n)^T$$

- the configuration space  $\mathcal{Q}$  is in general an  $n$ -dimensional smooth manifold, locally diffeomorphic to  $\mathbb{R}^n$
- the **generalized velocity** at a generic point of a trajectory  $q(t) \subset \mathcal{Q}$  is the tangent vector

$$\dot{q} = (\dot{q}_1 \quad \dot{q}_2 \quad \dots \quad \dot{q}_n)^T$$

- **geometric constraints** may exist or be imposed on the mechanical system

$$h_i(q) = 0 \quad i = 1, \dots, k$$

restricting the possible motions to an  $(n - k)$ -dimensional submanifold

- a mechanical system may also be subject to a set of **kinematic constraints**, involving generalized coordinates and their derivatives; e.g., first-order kinematic constraints

$$a_i(q, \dot{q}) = 0 \quad i = 1, \dots, k$$

- in most cases, the constraints are **Pfaffian**

$$a_i^T(q)\dot{q} = 0 \quad i = 1, \dots, k \quad \text{or} \quad A^T(q)\dot{q} = 0$$

i.e., they are linear in the velocities

- kinematic constraints may be **integrable**, that is, there may exist  $k$  functions  $h_i$  such that

$$\frac{\partial h_i(q(t))}{\partial q} = a_i^T(q) \quad i = 1, \dots, k$$

in this case, the kinematic constraints are indeed geometric constraints

a set of Pfaffian constraints is called **holonomic** if it is integrable (a geometric limitation); otherwise, it is called **nonholonomic** (a kinematic limitation)

holonomic/nonholonomic constraints affect mobility in a **completely different** way:

for illustration, consider a single Pfaffian constraint

$$a^T(q)\dot{q} = 0$$

- if the constraint is **holonomic**, then it can be integrated as

$$h(q) = c$$

with  $\frac{\partial h}{\partial q} = a^T(q)$  and  $c$  an integration constant



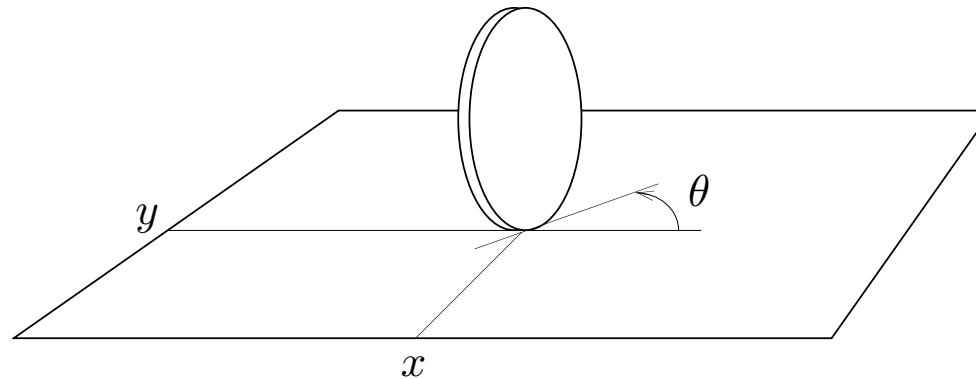
the motion of the system is confined to lie on a particular level surface (**leaf**) of  $h$ , depending on the initial condition through  $c = h(q_0)$

- if the constraint is **nonholonomic**, then it cannot be integrated



although at each configuration the instantaneous motion (velocity) of the system is restricted to an  $(n - 1)$ -dimensional space (the null space of the constraint matrix  $a^T(q)$ ), **it is still possible to reach any configuration in  $\mathcal{Q}$**

**a canonical example of nonholonomy:** the rolling disk

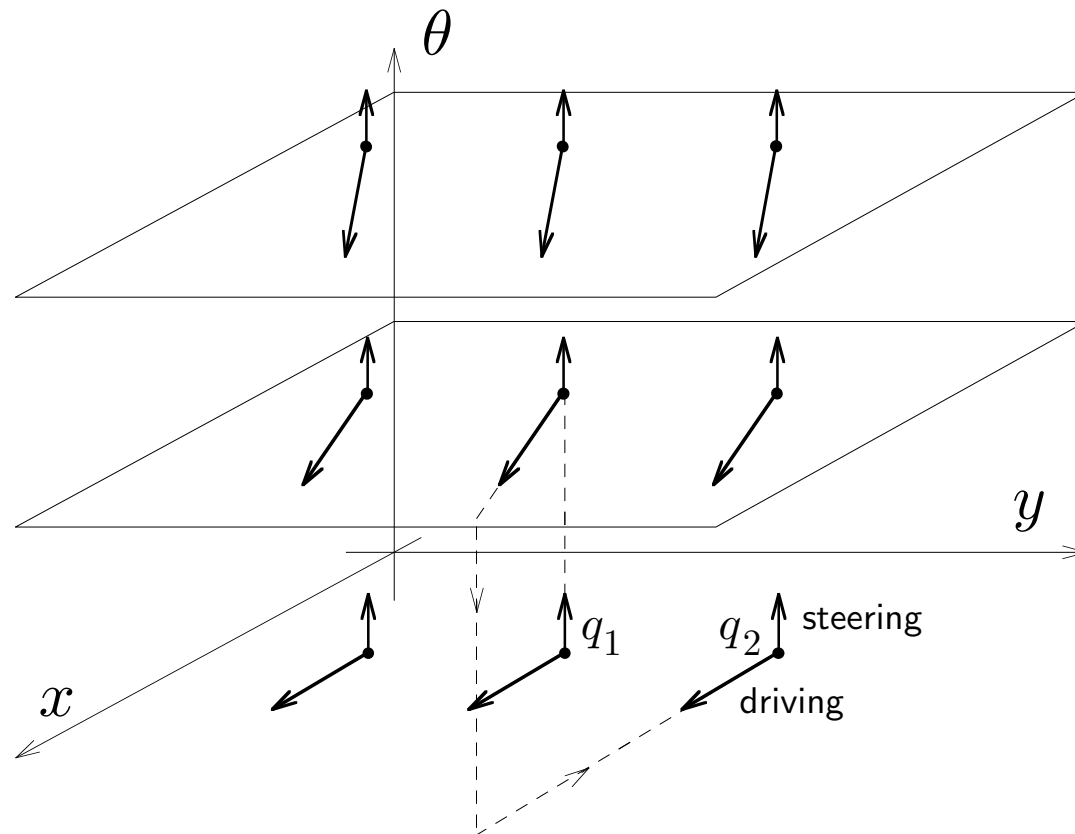


- generalized coordinates  $q = (x, y, \theta)$
- **pure rolling** nonholonomic constraint  $\dot{x} \sin \theta - \dot{y} \cos \theta = 0 \quad \left( \frac{\dot{y}}{\dot{x}} = \tan \theta \right)$
- feasible velocities are contained in the null space of the constraint matrix

$$a^T(q) = (\sin \theta \quad -\cos \theta \quad 0) \quad \Longrightarrow \quad \mathcal{N}(a^T(q)) = \text{span} \left\{ \begin{pmatrix} \cos \theta \\ \sin \theta \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \right\}$$

- **any** configuration  $q_f = (x_f, y_f, \theta_f)$  can be reached:
  1. rotate the disk until it aims at  $(x_f, y_f)$
  2. roll the disk until it reaches  $(x_f, y_f)$
  3. rotate the disk until its orientation is  $\theta_f$

nonholonomy **in the configuration space** of the rolling disk



- at each  $q$ , only two instantaneous directions of motion are possible
- to move from  $q_1$  to  $q_2$  (**parallel parking**) an appropriate **maneuver** (sequence of moves) is needed; one possibility is to follow the dashed line

a less canonical example of nonholonomy: the fifteen puzzle

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	

- generalized coordinates  $q = (q_1, \dots, q_{15})$
- each  $q_i$  may assume 16 different values corresponding to the cells in the grid; **legal** configurations are obtained when  $q_i \neq q_j$  for  $i \neq j$
- depending on the current configuration, a limited number (2 to 4) moves are possible
- **any** configuration with an **even** number of inversions can be reached by an appropriate sequence of moves

## Integrability of Kinematic Constraints

- when is a single kinematic Pfaffian constraint

$$a^T(q)\dot{q} = \sum_{j=1}^n a_j(q)\dot{q}_j = 0$$

integrable as  $h(q) = 0$ ?

since  $\dot{h}(q) = \sum_{j=1}^n \frac{\partial h}{\partial q_j} \dot{q}_j = 0$ , integrability requires

$$\gamma(q)a_j(q) = \frac{\partial h}{\partial q_j}(q) \quad j = 1, \dots, n$$

with  $\gamma(q) \neq 0$  **integrating factor**, or equivalently

$$\frac{\partial(\gamma a_k)}{\partial q_j} = \frac{\partial(\gamma a_j)}{\partial q_k}, \quad j, k = 1, \dots, n$$

where a system of PDE's must be solved

- for  $k$  kinematic Pfaffian constraints, one must check integrability **not only** of each constraint but **also** of independent combinations

$$\sum_{i=1}^k \gamma_i(q) a_i^T(q) \dot{q} = 0$$

even if each constraint is not integrable by itself, a subset (or even the whole set) of them may be integrable!

- if there exist  $p \leq k$  functions  $h_i$  such that,  $\forall q$

$$\text{span} \left\{ \frac{\partial h_1}{\partial q}(q), \dots, \frac{\partial h_p}{\partial q}(q) \right\} \subset \text{span} \{a_1^T(q), \dots, a_k^T(q)\}$$

then the system motion is restricted to the  $(n-p)$ -dimensional manifold of level surfaces of the  $h_i$ 's

$$\{q : h_1(q) = c_1, \dots, h_p(q) = c_p\}$$

- motion reduction due to kinematic constraints

$$\begin{aligned} p = k & \iff \text{holonomic} \\ 0 < p < k & \iff \text{partially holonomic} \\ p = 0 & \iff \text{(completely) nonholonomic} \end{aligned}$$

- assessing integrability is **not obvious**: complete (N&S conditions) and constructive answers are obtained by differential geometric tools

## A Control Viewpoint

- holonomy/nonholonomy of constraints may be conveniently studied through a dual approach: look at the

directions in which motion is **allowed**  
rather than  
directions in which motion is **prohibited**

- there is a strict relationship between  
capability of accessing every configuration  
and  
nonholonomy of the velocity constraints

- the interesting question is:

given two arbitrary points  $q_i$  and  $q_f$ ,  
when does a connecting trajectory  $q(t)$  exist  
which satisfies the kinematic constraints?



... this is indeed a **controllability** problem!

- associate to the set of kinematic constraints a basis for their null space, i.e. a set of vectors  $g_j$  such that

$$a_i^T(q)g_j(q) = 0 \quad i = 1, \dots, k \quad j = 1, \dots, n - k$$

or in matrix form

$$A^T(q)G(q) = 0$$

- feasible trajectories of the mechanical system are the solutions  $q(t)$  of

$$\dot{q} = \sum_{j=1}^m g_j(q)u_j = G(q)u \quad (*)$$

for some input  $u(t) \in \mathbb{R}^m$ ,  $m = n - k$  ( $u$ : also called **pseudovelocities**)

- (\*) is a **driftless** (i.e.,  $u=0 \Rightarrow \dot{q}=0$ ) nonlinear system known as the **kinematic model** of the constrained mechanical system
- **controllability** of its whole configuration space is equivalent to **nonholonomy** of the original kinematic constraints

## Dynamics versus Kinematics

- use Lagrange formalism to obtain the dynamics of a mechanical system with  $n$  degrees of freedom, subject to  $k$  Pfaffian kinematic constraints

$$A^T(q)\dot{q} = 0$$

- Lagrangian = Kinetic Energy – Potential Energy

$$\mathcal{L}(q, \dot{q}) = T(q, \dot{q}) - U(q) = \frac{1}{2} \dot{q}^T B(q) \dot{q} - U(q)$$

with inertia matrix  $B(q) > 0$

- **Euler-Lagrange** equations

$$\frac{d}{dt} \left( \frac{\partial \mathcal{L}}{\partial \dot{q}} \right)^T - \left( \frac{\partial \mathcal{L}}{\partial q} \right)^T = A(q)\lambda + S(q)\tau$$

where

- $S(q)$  is a  $n \times m$  matrix mapping the  $m$  external inputs  $\tau$  into forces/torques performing work on the generalized coordinates  $q$  ( $m = n - k$ )
- $\lambda \in \mathbb{R}^m$  is the vector of **Lagrange multipliers**

- the **dynamic model** of the mechanism subject to constraints is

$$\begin{aligned} B(q)\ddot{q} + n(q, \dot{q}) &= A(q)\lambda + S(q)\tau & (\diamond) \\ A^T(q)\dot{q} &= 0 \end{aligned}$$

with

$$n(q, \dot{q}) = \dot{B}(q)\dot{q} - \frac{1}{2} \left( \frac{\partial}{\partial q} (\dot{q}^T B(q) \dot{q}) \right)^T + \left( \frac{\partial U(q)}{\partial q} \right)^T$$

- to eliminate the Lagrange multipliers, being

$$A^T(q)G(q) = 0$$

multiply  $(\diamond)$  by  $G^T(q)$  to obtain a reduced set of  $m = n - k$  differential equations

$$G^T(q) (B(q)\ddot{q} + n(q, \dot{q})) = G^T(q)S(q)\tau$$

- assume now an hypothesis of 'enough control'

$$\det G^T(q)S(q) \neq 0$$

- merge the kinematic and dynamic models into the **reduced state-space model**

$$\begin{aligned}\dot{q} &= G(q)v \\ M(q)\dot{v} + m(q, v) &= G^T(q)S(q)\tau\end{aligned}$$

where  $v \in \mathbb{R}^m$  are the pseudovelocities and

$$\begin{aligned}M(q) &= G^T(q)B(q)G(q) > 0 \\ m(q, v) &= G^T(q)B(q)\dot{G}(q)v + G^T(q)n(q, G(q)v)\end{aligned}$$

where

$$\dot{G}(q)v = \sum_{i=1}^m \left( v_i \frac{\partial g_i}{\partial q} \right) G(q)v$$

- define external input  $\tau$  as a **nonlinear feedback law** from the state  $(q, v)$

$$\tau = (G^T(q)S(q))^{-1} (M(q)a + m(q, v)) \quad (\Delta)$$

where  $a \in \mathbb{R}^m$  is a vector of **pseudoaccelerations**

- in the absence of constraints,  $(\Delta)$  reduces to the **computed torque** law  $\Rightarrow$  linear & decoupled closed-loop dynamics (double integrators)

- due to the presence of constraints, the resulting system is

$$\begin{aligned}\dot{q} &= G(q)v && \text{kinematic model} \\ \dot{v} &= a && \text{dynamic extension}\end{aligned}$$

- letting  $x = (q, v)$  and  $a = u$ , the **state-space model** of the closed-loop system is rewritten in compact form as

$$\dot{x} = f(x) + g(x)u = \begin{pmatrix} G(q)v \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ I_m \end{pmatrix} u$$

i.e., a nonlinear control system **with drift** also known as the **second-order kinematic model** of the constrained mechanism



- ★ an invertible *feedback control law* can eliminate dynamic parameters
- ★ moving from kinematics to dynamics essentially requires some *input smoothness* assumptions (need  $a = \dot{v}$ )
- ★ most nonholonomic problems can be addressed at a *first-order* kinematic level

## More General Nonholonomic Constraints

- one may also find Pfaffian constraints of the form

$$a_i^T(q)\dot{q} = c_i, \quad i = 1, \dots, k \quad \text{or} \quad A^T(q)\dot{q} = c$$

with constant  $c_i$

- these constraints are **differential** but **not** of a kinematic nature; for example, this form arises from conservation of an initial **non-zero** angular momentum in space robots
- the mechanism subject to constraint is transformed into an equivalent control system by describing the feasible trajectories  $q(t)$  as solutions of

$$\dot{q} = f(q) + \sum_{i=1}^m g_i(q)u_i$$

i.e., a nonlinear control system **with drift**, where  $g_1(q), \dots, g_m(q)$  are a basis of the null space of  $A^T(q)$  and the drift vector  $f$  is computed through pseudoinversion

$$f(q) = A^\#(q)c = A(q) (A^T(q)A(q))^{-1} c$$

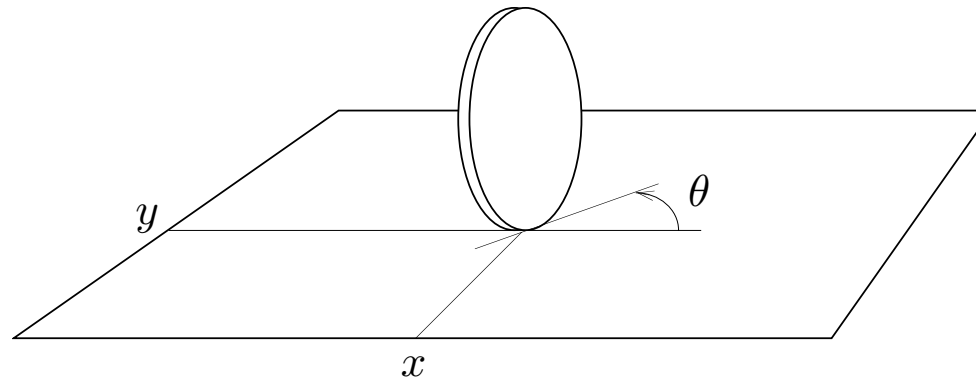
## MODELING EXAMPLES

source of nonholonomic constraints on motion:

- bodies in **rolling contact without slipping**
  - wheeled mobile robots (WMRs) or automobiles (wheels rolling on the ground with no skid or slippage)
  - dextrous manipulation with multifingered robot hands (rounded fingertips on grasped objects)
- **angular momentum conservation** in multibody systems
  - robotic manipulators floating in space (with no external actuation)
  - dynamically balancing hopping robots, divers or astronauts (in flying or mid-air phases)
  - satellites with reaction (or momentum) wheels for attitude stabilization
- special **control operation**
$$\dot{q} = G(q)u \quad q \in \mathbb{R}^n \quad u \in \mathbb{R}^m \quad (m < n)$$
  - non-cyclic inversion schemes for redundant robots ( $m$  task commands for  $n$  joints)
  - floating underwater robotic systems  
( $m = 4$  velocity inputs for  $n = 6$  generalized coords)

# Wheeled Mobile Robots

## unicycle



- generalized coordinates  $q = (x, y, \theta)$
- nonholonomic constraint  $\dot{x} \sin \theta - \dot{y} \cos \theta = 0$
- a matrix whose columns span the null space of the constraint matrix is

$$G(q) = \begin{pmatrix} \cos \theta & 0 \\ \sin \theta & 0 \\ 0 & 1 \end{pmatrix} = (g_1 \quad g_2)$$

- hence the kinematic model

$$\dot{q} = G(q)u = g_1(q)u_1 + g_2(q)u_2$$

with  $u_1 = \text{driving}$ ,  $u_2 = \text{steering}$  velocity inputs

## unicycle dynamics

- define

$m$  = mass of the unicycle

$I$  = inertia around vertical axis at contact point

$u_1$  = driving force

$u_2$  = steering torque

- the general dynamic model

$$B(q)\ddot{q} + n(q, \dot{q}) = a(q)\lambda + S(q)\tau$$

being  $B(q) = B$ ,  $n = 0$  particularizes in this case to

$$\begin{pmatrix} m & 0 & 0 \\ 0 & m & 0 \\ 0 & 0 & I \end{pmatrix} \begin{pmatrix} \ddot{x} \\ \ddot{y} \\ \ddot{\theta} \end{pmatrix} = \begin{pmatrix} \sin \theta \\ -\cos \theta \\ 0 \end{pmatrix} \lambda + \begin{pmatrix} \cos \theta & 0 \\ \sin \theta & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \tau_1 \\ \tau_2 \end{pmatrix}$$

subject to  $a^T(q)\dot{q} = 0$

from the reduction procedure, being

$$\begin{aligned}G(q) &= S(q) \\G^T(q)S(q) &= I_{2 \times 2} \\G^T(q)B\dot{G}(q) &= 0\end{aligned}$$

we obtain the reduced state-space model

$$\begin{aligned}\dot{q} &= G(q)v \\G^T(q)BG(q)\dot{v} &= \tau\end{aligned}$$

or the five dynamic equations

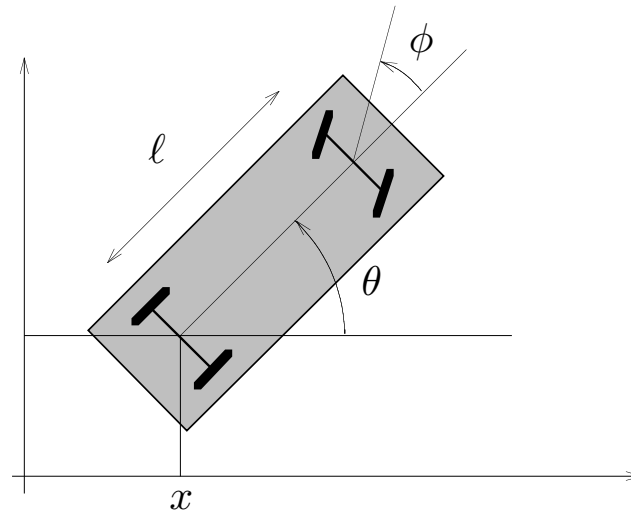
$$\begin{aligned}\dot{x} &= \cos \theta v_1 \\ \dot{y} &= \sin \theta v_1 \\ \dot{\theta} &= v_2 \\ m \dot{v}_1 &= \tau_1 \\ I \dot{v}_2 &= \tau_2\end{aligned}$$

that can be put in the form

$$\dot{X} = f(X) + g_1(X)\tau_1 + g_2(X)\tau_2$$

with  $X = (x, y, \theta, v_1, v_2)$

## car-like robot



- 'bicycle' model: front and rear wheels collapse into two wheels at the axle midpoints
- generalized coordinates  $q = (x, y, \theta, \phi)$      $\phi$ : steering angle
- nonholonomic constraints

$$\begin{aligned} \dot{x}_f \sin(\theta + \phi) - \dot{y}_f \cos(\theta + \phi) &= 0 && \text{(front wheel)} \\ \dot{x} \sin \theta - \dot{y} \cos \theta &= 0 && \text{(rear wheel)} \end{aligned}$$

- being the front wheel position

$$x_f = x + l \cos \theta \quad y_f = y + l \sin \theta$$

the first constraint becomes

$$\dot{x} \sin(\theta + \phi) - \dot{y} \cos(\theta + \phi) - \dot{\theta} l \cos \phi = 0$$

the constraint matrix is

$$A^T(q) = \begin{pmatrix} \sin(\theta + \phi) & -\cos(\theta + \phi) & -\ell \cos \phi & 0 \\ \sin \theta & -\cos \theta & 0 & 0 \end{pmatrix}$$

there are two physical alternatives for the controls:

(*RD*) choosing

$$G(q) = \begin{pmatrix} \cos \theta & 0 \\ \sin \theta & 0 \\ \frac{1}{\ell} \tan \phi & 0 \\ 0 & 1 \end{pmatrix} \implies \dot{q} = g_1(q)u_1 + g_2(q)u_2$$

where  $u_1 =$  **rear driving**,  $u_2 =$  **steering** inputs

◇ a ‘control singularity’ at  $\phi = \pm \pi/2$ , where vector field  $g_1$  diverges

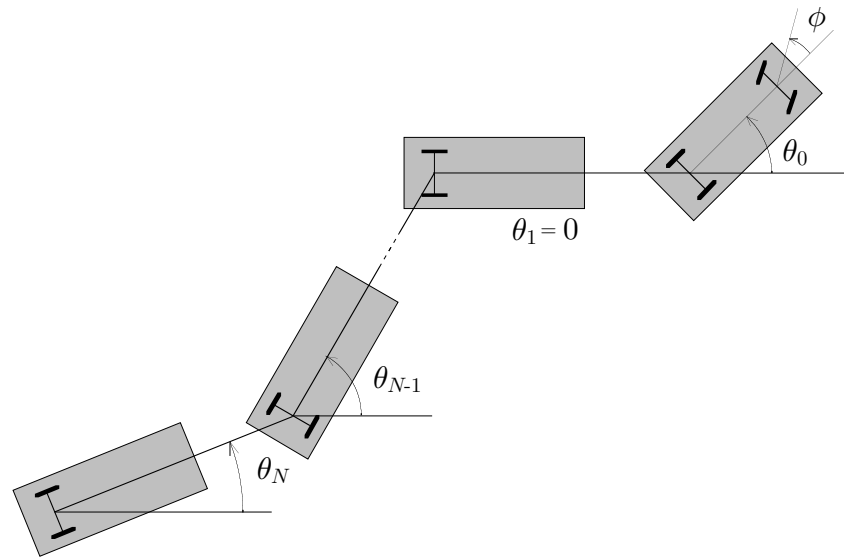
(*FD*) choosing

$$G(q) = \begin{pmatrix} \cos \theta \cos \phi & 0 \\ \sin \theta \cos \phi & 0 \\ \frac{1}{\ell} \sin \phi & 0 \\ 0 & 1 \end{pmatrix} \implies \dot{q} = g_1(q)u_1 + g_2(q)u_2$$

where  $u_1 =$  **front driving**,  $u_2 =$  **steering** inputs

◇ no singularities in this case!

## ***N*-trailer system**



- an FD car-like robot with  $N$  trailers, each hinged to the axle midpoint of the previous

- generalized coordinates  $q = (x, y, \phi, \theta_0, \theta_1, \dots, \theta_N) \in \mathbb{R}^{N+4}$

$x, y$  = position of the car rear axle midpoint

$\phi$  = steering angle of the car (w.r.t. car body)

$\theta_0$  = orientation angle of the car (w.r.t.  $x$ -axis)

$\theta_i$  = orientation angle of  $i$ -th trailer (w.r.t.  $x$ )

- the car is considered as the 0-th trailer

$d_0 = \ell =$  car length

$d_i =$   $i$ -th trailer length (hinge to hinge)

nonholonomic constraints:

### steering wheel

$$\dot{x}_f \sin(\theta_0 + \phi) - \dot{y}_f \cos(\theta_0 + \phi) = 0$$

with

$$x_f = x + \ell \cos \theta_0 \quad y_f = y + \ell \sin \theta_0$$

### all other wheels

$$\dot{x}_i \sin \theta_i - \dot{y}_i \cos \theta_i = 0 \quad i = 0, 1, \dots, N$$

being

$$x_i = x - \sum_{j=1}^i d_j \cos \theta_j \quad y_i = y - \sum_{j=1}^i d_j \sin \theta_j$$

the constraints become

$$\begin{aligned} \dot{x} \sin(\theta_0 + \phi) - \dot{y} \cos(\theta_0 + \phi) - \dot{\theta}_0 \ell \cos \phi &= 0 \\ \dot{x} \sin \theta_i - \dot{y} \cos \theta_i + \sum_{j=1}^i \dot{\theta}_j d_j \cos(\theta_i - \theta_j) &= 0 \quad i = 0, 1, \dots, N \end{aligned}$$

- the null space of the  $N + 2$  constraints is spanned by the two columns  $g_1, g_2$  of

$$G(q) = \begin{pmatrix} \cos \theta_0 & 0 \\ \sin \theta_0 & 0 \\ 0 & 1 \\ \frac{1}{\ell} \tan \phi & 0 \\ -\frac{1}{d_1} \sin(\theta_1 - \theta_0) & 0 \\ -\frac{1}{d_2} \cos(\theta_1 - \theta_0) \sin(\theta_2 - \theta_1) & 0 \\ \vdots & \vdots \\ -\frac{1}{d_i} \left( \prod_{j=1}^{i-1} \cos(\theta_j - \theta_{j-1}) \right) \sin(\theta_i - \theta_{i-1}) & 0 \\ \vdots & \vdots \\ -\frac{1}{d_N} \left( \prod_{j=1}^{N-1} \cos(\theta_j - \theta_{j-1}) \right) \sin(\theta_N - \theta_{N-1}) & 0 \end{pmatrix}$$

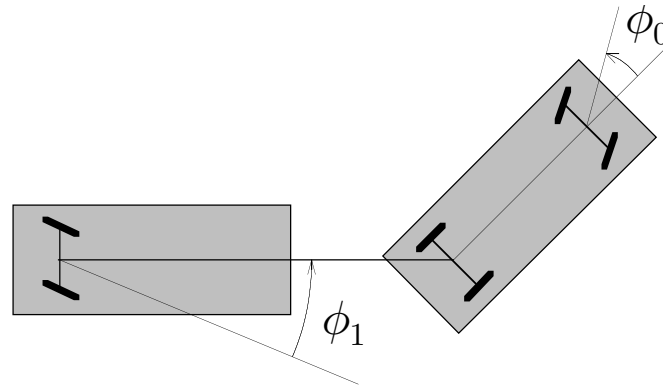
- the kinematic model is  $\dot{q} = g_1(q)u_1 + g_2(q)u_2$   
with  $u_1 = \text{(rear) driving}$ ,  $u_2 = \text{steering}$  inputs for the front car
- an alternative way to derive kinematic equations

$$\begin{aligned} \dot{\theta}_i &= -\frac{1}{d_i} \sin(\theta_i - \theta_{i-1}) \nu_{i-1} \\ \nu_i &= \nu_{i-1} \cos(\theta_i - \theta_{i-1}) \end{aligned} \quad i = 1, \dots, N$$

with  $\nu_i =$  linear (forward) velocity of the  $i$ -th trailer ( $\nu_0 = u_1$ )

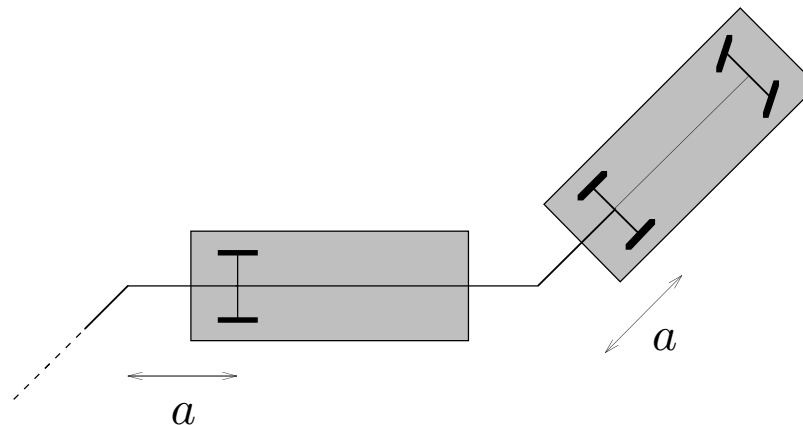
## other wheeled mobile robots

- **firetruck**



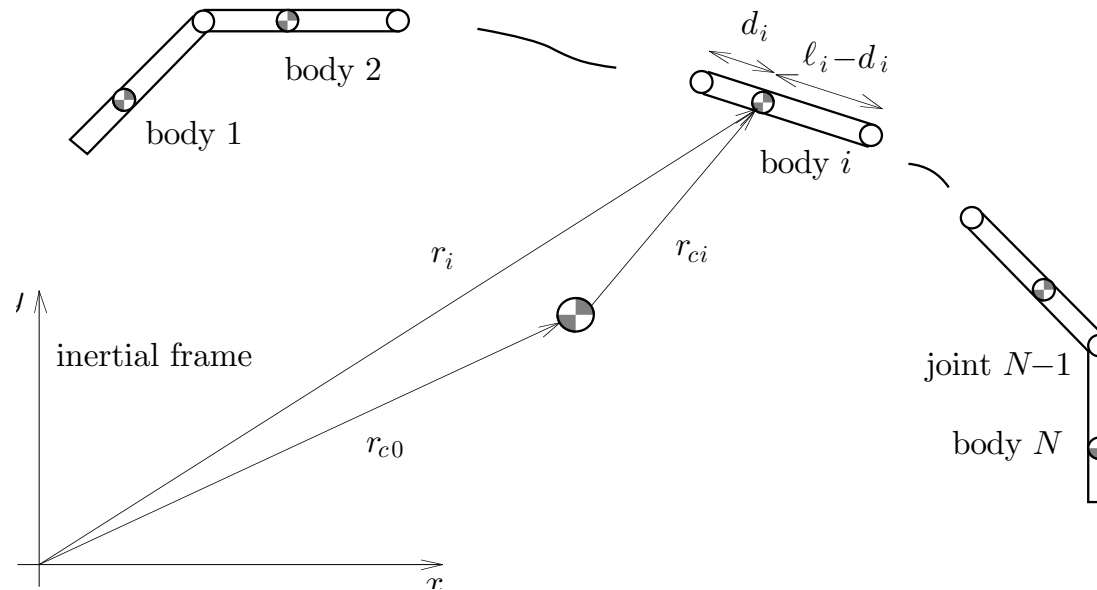
6 configuration variables, 3 differential constraints, 3 control inputs (car driving and steering, trailer steering)

- $N$ -trailer system with **nonzero hooking**



when  $a \neq 0$  and  $N \geq 2$ , this system **cannot** be converted in chained form (later)

## Space Robots with Planar Structure



- $N$  planar bodies actuated at the  $N - 1$  joints (**internal** forces only)

- for the  $i$ -th body, let:

$m_i, I_i$  = mass and inertia matrix

$r_i, v_i$  = position and velocity of the center of mass

$\omega_i$  = angular velocity

- assume the center of mass of each body is located on the body axis

- no external forces (gravity), no dissipation

⇓

1. conservation of linear momentum (assumed to be initially zero)

$$\sum_{i=1}^N m_i v_i = 0 \quad \Rightarrow \quad \sum_{i=1}^N m_i r_i = m_t r_{c0}$$

i.e., two scalar **holonomic** constraints in the planar case

2. conservation of angular momentum (= zero)

$$\sum_{i=1}^N (I_i \omega_i + m_i (r_i \times v_i)) = 0$$

i.e., a scalar **nonholonomic** constraint in the planar case

- it is convenient to place the inertial frame in the center of mass of the whole system ( $r_{c0} = 0$ ,  $r_i = r_{ci}$ )
- for a  $N$ -body system with kinetic energy

$$T = \frac{1}{2} \dot{q}^T B(q) \dot{q}$$

and  $U = \text{constant}$ , the vector of **generalized momenta** is

$$p = B(q) \dot{q} \in \mathbb{R}^N$$

- for a planar system, each component

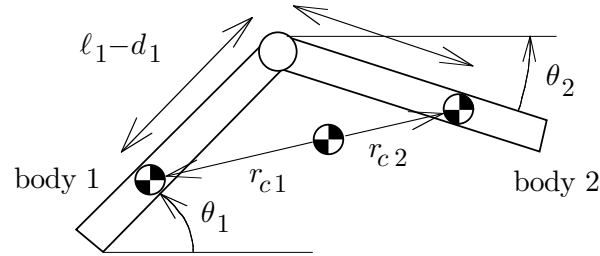
$$p_i = b_i^T(q) \dot{q}$$

represents an angular momentum along the  $z$  axis (orthogonal to the  $xy$  plane); thus, conservation of (zero) angular momentum can be expressed as a Pfaffian constraint:

$$\sum_{i=1}^N p_i = \sum_{i=1}^N b_i^T(q) \dot{q} = \mathbf{1}^T B(q) \dot{q} = A^T(q) \dot{q} = 0,$$

where  $\mathbf{1} = (1, 1, \dots, 1)$

## 2-body space robot



from the two vector equations

$$\begin{pmatrix} r_{c1x} \\ r_{c1y} \end{pmatrix} + (\ell_1 - d_1) \begin{pmatrix} \cos \theta_1 \\ \sin \theta_1 \end{pmatrix} + d_2 \begin{pmatrix} \cos \theta_2 \\ \sin \theta_2 \end{pmatrix} = \begin{pmatrix} r_{c2x} \\ r_{c2y} \end{pmatrix}$$

$$m_1 \begin{pmatrix} r_{c1x} \\ r_{c1y} \end{pmatrix} + m_2 \begin{pmatrix} r_{c2x} \\ r_{c2y} \end{pmatrix} = 0$$

one solves for

$$\begin{pmatrix} r_{c1} \\ r_{c2} \end{pmatrix} = \begin{pmatrix} r_{c1x} \\ r_{c1y} \\ r_{c2x} \\ r_{c2y} \end{pmatrix} = \begin{pmatrix} k_{11} \begin{pmatrix} \cos \theta_1 \\ \sin \theta_1 \end{pmatrix} + k_{12} \begin{pmatrix} \cos \theta_2 \\ \sin \theta_2 \end{pmatrix} \\ k_{21} \begin{pmatrix} \cos \theta_1 \\ \sin \theta_1 \end{pmatrix} + k_{22} \begin{pmatrix} \cos \theta_2 \\ \sin \theta_2 \end{pmatrix} \end{pmatrix}$$

where (setting  $m_t = m_1 + m_2$ )

$$k_{11} = -\frac{m_2(\ell_1 - d_1)}{m_t} \quad k_{12} = -\frac{m_2 d_2}{m_t}$$

$$k_{21} = \frac{m_1(\ell_1 - d_1)}{m_t} \quad k_{22} = \frac{m_1 d_2}{m_t}$$

- kinetic energy of the system  $T = T_1 + T_2$ , with

$$T_i = \frac{1}{2} m_i \dot{r}_{ci}^T \dot{r}_{ci} + \frac{1}{2} I_{zzi} \dot{\theta}_i^2 \quad i = 1, 2$$

so that

$$T = \frac{1}{2} \begin{pmatrix} \dot{\theta}_1 & \dot{\theta}_2 \end{pmatrix} \begin{pmatrix} \bar{I}_1 & b_{12}(\theta_2 - \theta_1) \\ b_{12}(\theta_2 - \theta_1) & \bar{I}_2 \end{pmatrix} \begin{pmatrix} \dot{\theta}_1 \\ \dot{\theta}_2 \end{pmatrix}$$

where

$$\begin{aligned} \bar{I}_i &= I_{zzi} + m_1 k_{1i}^2 + m_2 k_{2i}^2 \quad i = 1, 2 \\ b_{12} &= (m_1 k_{11} k_{12} + m_2 k_{21} k_{22}) \cos(\theta_2 - \theta_1) \end{aligned}$$

- since  $T$  is only a function of  $\phi_1 = \theta_2 - \theta_1$ , the conservation of momentum can be written as the differential constraint

$$\begin{pmatrix} 1 & 1 \end{pmatrix} \begin{pmatrix} \bar{I}_1 & b_{12}(\phi_1) \\ b_{12}(\phi_1) & \bar{I}_2 \end{pmatrix} \left( \begin{pmatrix} 1 \\ 1 \end{pmatrix} \dot{\theta}_1 + \begin{pmatrix} 0 \\ 1 \end{pmatrix} \dot{\phi}_1 \right) = 0$$

from which

$$\dot{\theta}_1 = -\frac{\bar{I}_2 + b_{12}(\phi_1)}{\bar{I}_1 + \bar{I}_2 + 2b_{12}(\phi_1)} \dot{\phi}_1$$

- taking the single joint velocity  $\dot{\phi}_1 = u$  as input and using as generalized coordinates

$$q = \begin{pmatrix} \theta_1 \\ \phi_1 \end{pmatrix} \quad \begin{array}{l} \text{base angle (absolute orientation)} \\ \text{relative angle (shape)} \end{array}$$

the kinematic model describing all the system feasible trajectories is

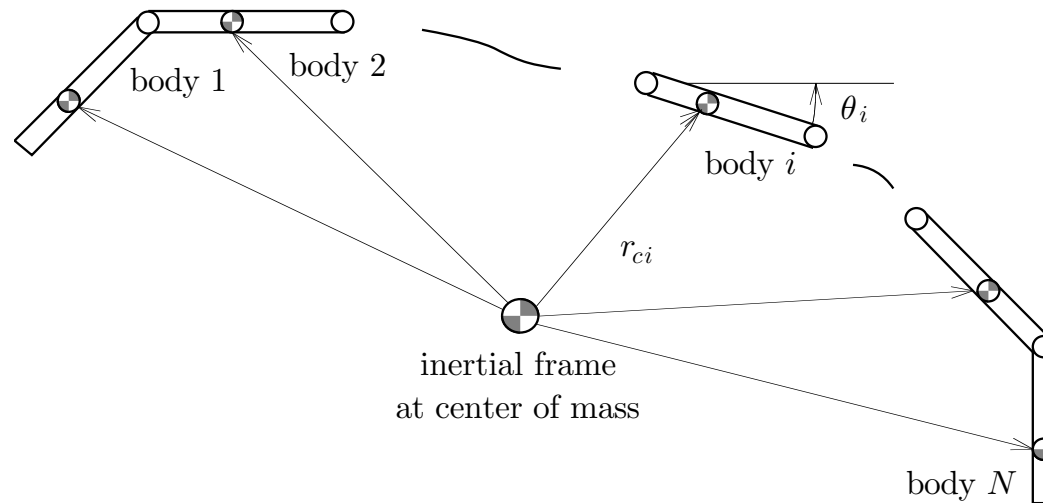
$$\dot{q} = g(q)u = \begin{pmatrix} -\frac{\bar{I}_2 + b_{12}(\phi_1)}{\bar{I}_1 + \bar{I}_2 + 2b_{12}(\phi_1)} \\ 1 \end{pmatrix} u$$

- it may be shown (see later) that such system is **not controllable**; thus, the constraint expressing conservation of the angular momentum is in this case **integrable**  
in particular, if  $\bar{I}_1 = \bar{I}_2$

$$\dot{\theta}_1 = -\frac{1}{2}\dot{\phi}_1 \quad \Rightarrow \quad \theta_1 = -\frac{1}{2}\phi_1 + k$$

- angular momentum conservation is a **holonomic** constraint for a planar space robot with  $N = 2$  bodies
- this mechanical system **cannot be controlled** through  $u$  so as to achieve **an arbitrary pair** of absolute orientation **and** internal shape

## *N*-body space robot



- follow the same steps as before, with the inertial reference frame placed at the system center of mass and  $\theta_i =$  absolute angle of  $i$ -th body
- position of center of mass of  $i$ -th body

$$\begin{pmatrix} r_{cix} \\ r_{ciy} \end{pmatrix} = \begin{pmatrix} \sum_{j=1}^N k_{ij} \cos \theta_j \\ \sum_{j=1}^N k_{ij} \sin \theta_j \end{pmatrix}$$

where

$$k_{ij} = \begin{cases} \frac{1}{m_i} \left( \ell_j \sum_{h=1}^{j-1} m_h + (\ell_j - d_j) m_j \right) & (j < i) \\ \frac{1}{m_i} \left( d_i \sum_{h=1}^{i-1} m_h - (\ell_i - d_i) \sum_{k=i+1}^N m_k \right) & (j = i) \\ \frac{1}{m_i} \left( -\ell_j \sum_{h=j+1}^N m_h - d_j m_j \right) & (j > i) \end{cases}$$

- kinetic energy of  $i$ -th body

$$\begin{aligned}
 T_i &= \frac{1}{2} m_i \dot{r}_{ci}^T \dot{r}_{ci} + \frac{1}{2} I_{zzi} \dot{\theta}_i^2 \\
 &= \frac{1}{2} m_i \left( \sum_{h=1}^N \sum_{j=1}^N k_{ij} k_{ih} \cos(\theta_h - \theta_j) \dot{\theta}_h \dot{\theta}_j \right) + \frac{1}{2} I_{zzi} \dot{\theta}_i^2
 \end{aligned}$$

- kinetic energy of the system

$$T = \sum_{i=1}^N T_i = \frac{1}{2} \dot{\theta}^T B(\theta) \dot{\theta}$$

with elements of inertia matrix  $B = \{b_{ij}(\theta_i - \theta_j)\}$

$$b_{ij} = \begin{cases} \sum_{h=1}^N m_h k_{hi} k_{hj} \cos(\theta_i - \theta_j) & i \neq j \\ \sum_{h=1}^N m_h k_{hh}^2 + I_{zzi} & i = j \end{cases}$$

depending only on **relative** angles between bodies

- let

$$\begin{aligned}
 \phi_i &= \theta_{i+1} - \theta_i & i = 1, \dots, N-1 \\
 \Rightarrow & \phi = P\theta
 \end{aligned}$$

where  $P$  is a  $(N-1) \times N$  matrix

- redefine generalized coordinates as  $q = (\theta_1, \phi)$

$$q = \begin{pmatrix} 1 & \mathbf{0}^T \\ & P \end{pmatrix} \theta = \begin{pmatrix} 1 & 0 & 0 & \dots & \dots \\ -1 & 1 & 0 & \dots & \dots \\ 0 & -1 & 1 & 0 & \dots \\ \dots & \dots & \dots & -1 & 1 \end{pmatrix} \theta$$

with the inverse mapping

$$\theta = \begin{pmatrix} 1 & 0 & 0 & \dots & \dots \\ 1 & 1 & 0 & \dots & \dots \\ 1 & 1 & 1 & 0 & \dots \\ \dots & \dots & \dots & \dots & \dots \\ 1 & \dots & 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} \theta_1 \\ \phi \end{pmatrix} = \begin{pmatrix} 1 & S \end{pmatrix} \begin{pmatrix} \theta_1 \\ \phi \end{pmatrix}$$

where  $S$  is a  $N \times (N - 1)$  matrix

- conservation of angular momentum becomes

$$\mathbf{1}^T B(\phi) (\mathbf{1}\dot{\theta}_1 + S\dot{\phi}) = 0$$

from which

$$\dot{\theta}_1 = -\frac{\mathbf{1}^T B(\phi) S}{\mathbf{1}^T B(\phi) \mathbf{1}} v$$

where  $\dot{\phi} = v$  are the robot joint velocities

- the **kinematic model** of the  $N$ -body space robot is then

$$\dot{q} = \begin{pmatrix} \dot{\theta}_1 \\ \dot{\phi} \end{pmatrix} = \begin{pmatrix} s_1(\phi) & s_2(\phi) & \dots & s_{N-1}(\phi) \\ & \mathbf{I}_{N-1} & & \end{pmatrix} v$$

in which

$$s_i(\phi) = -\frac{s'_i(\phi)}{\mathbf{1}^T B(\phi) \mathbf{1}}$$

where the positive denominator is given by

$$\sum_{j=1}^N \bar{I}_j + \sum_{j=1}^N \sum_{\substack{h=1 \\ h \neq j}}^N \sum_{l=1}^N m_l k_{lj} k_{lh} \cos \left( \sum_{r=h}^{j-1} \phi_r \right)$$

with

$$\bar{I}_j = I_{zzj} + \sum_{h=1}^N m_h k_{hj}^2$$

and the numerator is

$$s'_i(\phi) = \sum_{j=i+1}^N \left( \bar{I}_j + \sum_{h=1}^N \sum_{l=1}^N m_l k_{lj} k_{lh} \cos \left( \sum_{r=h}^{j-1} \phi_r \right) \right)$$

e.g., in the case of  $N = 3$  bodies

$$\begin{aligned}s'_1 &= \bar{I}_2 + \bar{I}_3 + h_{12} \cos \phi_1 + 2h_{23} \cos \phi_2 + h_{13} \cos(\phi_1 + \phi_2) \\ s'_2 &= \bar{I}_3 + h_{23} \cos \phi_2 + h_{13} \cos(\phi_1 + \phi_2)\end{aligned}$$

and

$$\mathbf{1}^T B(\phi) \mathbf{1} = \bar{I}_1 + \bar{I}_2 + \bar{I}_3 + 2(h_{12} \cos \phi_1 + h_{23} \cos \phi_2 + h_{13} \cos(\phi_1 + \phi_2))$$

with

$$\begin{aligned}\bar{I}_1 &= m_1 k_{11}^2 + m_2 k_{21}^2 + m_3 k_{31}^2 + I_{zz1} \\ \bar{I}_2 &= m_1 k_{12}^2 + m_2 k_{22}^2 + m_3 k_{32}^2 + I_{zz2} \\ \bar{I}_3 &= m_1 k_{13}^2 + m_2 k_{23}^2 + m_3 k_{33}^2 + I_{zz3} \\ h_{12} &= m_1 k_{11} k_{12} + m_2 k_{21} k_{22} + m_3 k_{31} k_{32} \\ h_{13} &= m_1 k_{11} k_{13} + m_2 k_{21} k_{23} + m_3 k_{31} k_{33} \\ h_{23} &= m_1 k_{12} k_{13} + m_2 k_{22} k_{23} + m_3 k_{32} k_{33}\end{aligned}$$

with the  $k_{ij}$ 's and  $m_t$  depending on the inertial parameters

- the dynamic model of the  $N$ -body space robot is

$$B(\theta)\ddot{\theta} + n(\theta, \dot{\theta}) = P^T \tau$$

where  $\tau =$  **torques** at the  $N-1$  robot joints, with

$$\mathbf{1}^T B(\theta) \dot{\theta} = 0$$

- the **reduced dynamic model** (in the 'shape space') consists of  $2N - 1$  first-order differential equations

$$\begin{aligned} \dot{\theta}_1 &= -\frac{\mathbf{1}^T B(\phi) S}{\mathbf{1}^T B(\phi) \mathbf{1}} v \\ \dot{\phi} &= v \\ \dot{v} &= M^{-1}(\phi) (-m(\phi, v) + \tau) \end{aligned}$$

where

$$\begin{aligned} M(\phi) &= PB(\phi)P^T \\ m(\phi, v) &= \dot{M}(\phi)v - \frac{1}{2} \frac{\partial}{\partial \phi} (v^T M(\phi)v) \end{aligned}$$

- the right hand side of the above is **independent** of  $\theta_1$

↓

in this case, the mechanical system is referred to as a nonholonomic **Čaplygin** system

## TOOLS FROM DIFFERENTIAL GEOMETRY

- a smooth **vector field**  $f : \mathbb{R}^n \mapsto T_q\mathbb{R}^n$  is a smooth mapping from each point of  $\mathbb{R}^n$  to the tangent space  $T_q\mathbb{R}^n$

- if  $f$  defines the rhs of a differential equation

$$\dot{q} = f(q)$$

the **flow**  $\phi_t^f(q)$  of the vector field  $f$  is the mapping which associates to each  $q$  the solution evolving from  $q$ , i.e., it satisfies

$$\frac{d}{dt} \phi_t^f(q) = f(\phi_t^f(q))$$

with the **group** property  $\phi_t^f \circ \phi_s^f = \phi_{t+s}^f$

in linear systems,  $f(q) = Aq$ , the flow is  $\phi_t^f = e^{At}$

- considering two vector fields  $g_1$  and  $g_2$  as in

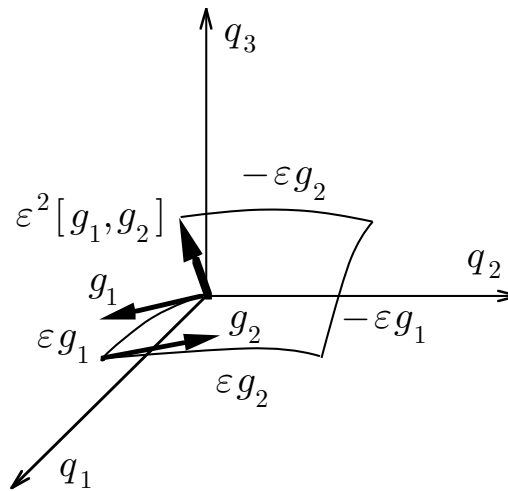
$$\dot{q} = g_1(q)u_1 + g_2(q)u_2$$

the composition of their flows (obtained by taking  $u_1 = \{1, 0\}$  and  $u_2 = \{0, 1\}$ ) is **non-commutative**

$$\phi_t^{g_1} \circ \phi_s^{g_2} \neq \phi_s^{g_2} \circ \phi_t^{g_1}$$

- starting at  $q_0$ , an infinitesimal flow of time  $\epsilon$  along  $g_1$ , then  $g_2$ , then  $-g_1$ , and finally  $-g_2$ , yields (R. Brockett: 'a computation everybody should do once in his life')

$$q(4\epsilon) = \phi_{\epsilon}^{-g_2} \circ \phi_{\epsilon}^{-g_1} \circ \phi_{\epsilon}^{g_2} \circ \phi_{\epsilon}^{g_1}(q_0) = q_0 + \epsilon^2 \left( \frac{\partial g_2}{\partial q} g_1(q_0) - \frac{\partial g_1}{\partial q} g_2(q_0) \right) + O(\epsilon^3)$$



- Lie bracket** of two vector fields  $g_1, g_2$

$$[g_1, g_2](q) = \frac{\partial g_2}{\partial q} g_1(q) - \frac{\partial g_1}{\partial q} g_2(q)$$

- $g_1$  and  $g_2$  **commute** if  $[g_1, g_2] = 0$ ; moreover,

$$[g_1, g_2] = 0 \quad \Rightarrow \quad q(4\epsilon) = q_0 \quad (\text{zero net flow})$$

- **properties** of Lie brackets

$$[f, g] = -[g, f]$$

**skew-symmetry**

$$[f, [g, h]] + [h, [f, g]] + [g, [h, f]] = 0$$

**Jacobi identity**

and the **chain rule**

$$[\alpha f, \beta g] = \alpha\beta[f, g] + \alpha(L_f\beta)g - \beta(L_g\alpha)f$$

with  $\alpha, \beta: \mathbb{R}^n \mapsto \mathbb{R}$  and the **Lie derivative** of  $\alpha$  along  $g$  defined as

$$L_g\alpha(q) = \frac{\partial\alpha}{\partial q}g(q)$$

in linear single input systems,  $f(q) = Aq$ ,  $g(q) = b$ ,

$$\begin{aligned} [f, g] &= -Ab & [f, [f, g]] &= A^2b \\ [f, [f, [f, g]]] &= -A^3b & \dots & \end{aligned}$$

- a smooth **distribution**  $\Delta$  associated with a set of smooth vector fields  $\{g_1, \dots, g_m\}$  assigns to each point  $q$  a subspace of its tangent space defined as

$$\begin{aligned} \Delta &= \text{span}\{g_1, \dots, g_m\} \\ &\Downarrow \\ \Delta_q &= \text{span}\{g_1(q), \dots, g_m(q)\} \subset T_q\mathbb{R}^n \end{aligned}$$

- a distribution is **regular** if  $\dim \Delta_q = \text{const}, \forall q$
- a distribution is **involutive** if it is closed under the Lie bracket operation
 
$$\Delta \text{ involutive} \iff \forall g_i, g_j \in \Delta \quad [g_i, g_j] \in \Delta$$
- the **involutive closure**  $\bar{\Delta}$  of a distribution  $\Delta$  is its closure under the Lie bracket operation
- the set of smooth vector fields on  $\mathbb{R}^n$  with the Lie bracket operation is a **Lie algebra**
- a Lie algebra is **nilpotent** if all Lie brackets of order  $\geq k$  (finite integer) vanish
- a regular distribution  $\Delta$  on  $\mathbb{R}^n$  of dimension  $k$  is **integrable** when there exist  $n - k$  independent functions  $h_i$  such that,  $\forall q$  and  $\forall g_j \in \Delta$

$$L_{g_j} h_i = \frac{\partial h_i}{\partial q} g_j(q) = 0 \quad i = 1, \dots, n - k$$

- the hypersurfaces defined as the level sets

$$\{q : h_1(q) = c_1, \dots, h_{n-k}(q) = c_{n-k}\}$$

are **integral manifolds** of  $\Delta$

## Frobenius Theorem

*a regular distribution is integrable if and only if it is involutive*

- $\Rightarrow$  a distribution of dimension 1 (i.e., associated to a single vector field) is **always** integrable
- the proof of sufficiency is constructive
- if the distribution  $\Delta$  of dimension  $k$  is involutive, then its integral manifolds (level sets of functions  $h_i$ ) are **leaves** of a **foliation** of  $\mathbb{R}^n$

**e.g.** the distribution  $\Delta = \text{span}\{g_1, g_2\}$  with

$$g_1(q) = \begin{pmatrix} 1 \\ q_2 \\ 0 \end{pmatrix} \quad g_2(q) = \begin{pmatrix} 1 \\ 0 \\ q_3 \end{pmatrix}$$

is involutive, because

$$[g_1, g_2](q) = 0$$

it induces a foliation of  $\mathbb{R}^3$  according to

$$q_1 - \log(q_2 q_3) = c \quad c \in \mathbb{R}$$

## Integrability of Pfaffian Constraints

- a smooth **one-form** is a mapping  $a^T: \mathbb{R}^n \mapsto T_q^* \mathbb{R}^n$ , the dual space of linear forms on  $T_q \mathbb{R}^n$

NB: one forms are represented in local coordinates as **row vectors** (hence the transpose notation!)

$$a^T(q) = (a_1(q) \quad a_2(q) \quad \dots \quad a_n(q))$$

- an **exact one-form**  $\omega^T$  is the differential of a smooth function  $h$

$$\omega^T = \frac{\partial h}{\partial q} = \left( \frac{\partial h}{\partial q_1} \quad \frac{\partial h}{\partial q_2} \quad \dots \quad \frac{\partial h}{\partial q_n} \right)$$

- a smooth **codistribution**  $A^T$  assigns to each point  $q$  a subspace of the dual of its tangent space and can be defined by a set of smooth one-forms  $a_i^T$

$$\begin{aligned} A^T &= \text{span} \{a_1^T, \dots, a_k^T\} \\ &\Updownarrow \\ A_q^T &= \text{span} \{a_1^T(q), \dots, a_k^T(q)\} \subset T_q^* \mathbb{R}^n \end{aligned}$$

- **distribution annihilating a codistribution**

given a set of smooth independent one-forms

$$a_i^T(q) \quad i = 1, \dots, k$$

which define a codistribution  $A^T$ , there exist smooth independent vector fields

$$g_j(q) \quad j = 1, \dots, n - k = m$$

defining a distribution  $\Delta = (A^T)^\perp$  such that

$$a_i^T(q) \cdot g_j(q) = 0 \quad \forall i, j$$

i.e., distribution  $\Delta$  annihilates codistribution  $A^T$



A set of Pfaffian constraints is integrable  
if and only its annihilating distribution is involutive

# CONTROL PROPERTIES

## Controllability of Nonholonomic Systems

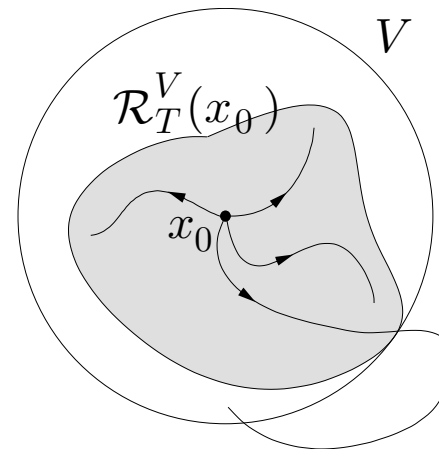
consider a nonlinear control system

$$\dot{x} = f(x) + \sum_{j=1}^m g_j(x)u_j \quad (\text{NCS})$$

with state  $x \in \mathcal{M} \simeq \mathbb{R}^n$ , and input in the class  $\mathcal{U}$  of piecewise-continuous time functions

- denote its unique solution at time  $t \geq 0$  by  $x(t, 0, x_0, u)$ , with input  $u(\cdot)$ , and  $x(0) = x_0$
- (NCS) is **controllable** if  $\forall x_1, x_2 \in \mathcal{M}, \exists T < \infty, \exists u: [0, T] \rightarrow \mathcal{U} : x(T, 0, x_1, u) = x_2$
- the set of states **reachable** from  $x_0$  **within** time  $T > 0$ , with trajectories contained in a neighborhood  $V$  of  $x_0$ , is denoted by

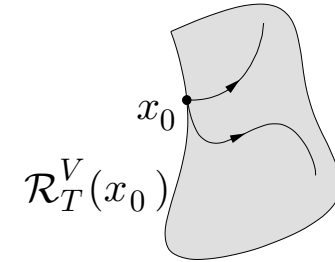
$$\mathcal{R}_T^V(x_0) = \bigcup_{\tau \leq T} \mathcal{R}^V(x_0, \tau)$$



where  $\mathcal{R}^V(x_0, \tau) = \{x \in \mathcal{M} \mid x(\tau, 0, x_0, u) = x, \forall t \in [0, \tau], x(t, 0, x_0, u) \in V\}$

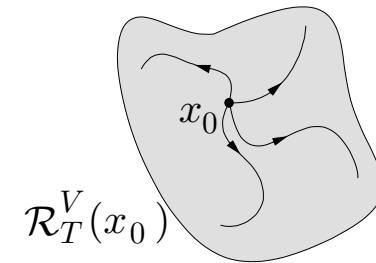
- (NCS) is **locally accessible** (LA) from  $x_0$  if  $\forall V$ , a neighborhood of  $x_0$ , and  $\forall T > 0$

$$\mathcal{R}_T^V(x_0) \supset \Omega, \quad \text{with } \Omega \text{ some non-empty open set}$$



- (NCS) is **small-time locally controllable** (STLC) from  $x_0$  if  $\forall V$ , a neighborhood of  $x_0$ , and  $\forall T > 0$

$$\mathcal{R}_T^V(x_0) \supset \Psi, \quad \text{with } \Psi \text{ some neighborhood of } x_0$$



- STLC  $\Rightarrow$  controllability  $\Rightarrow$  LA (not vice versa)
- LA is checked through an algebraic test
  - let  $\bar{\mathcal{C}}$  be the involutive closure of the distribution associated with  $\{f, g_1, g_2, \dots, g_m\}$
  - **Chow Theorem** (1939): (NCS) is LA from  $x_0$  if and only if
 
$$\dim \bar{\mathcal{C}}(x_0) = n \quad \text{accessibility rank condition}$$
  - an algorithmic test:

$$\bar{\mathcal{C}} = \text{span} \left\{ v \in \bigcup_{k \geq 0} \mathcal{C}^k \right\} \quad \text{with} \quad \begin{cases} \mathcal{C}^0 = \text{span} \{f, g_1, \dots, g_m\} \\ \mathcal{C}^k = \mathcal{C}^{k-1} + \text{span} \{[f, v], [g_j, v], j = 1, \dots, m : v \in \mathcal{C}^{k-1}\} \end{cases}$$

- only **sufficient** conditions exists for STLC , e.g., [Sussmann 1987]
- however, for driftless control systems:

$$\text{LA} \iff \text{controllability} \iff \text{STLC}$$

- this equivalence holds also whenever

$$f(x) \in \text{span} \{g_1(x), \dots, g_m(x)\} \quad \forall x \in \mathcal{M}$$

(‘trivial’ drift)

- if the driftless control system

$$\dot{x} = \sum_{i=1}^m g_i(x) u_i$$

is controllable, then its **dynamic extension**

$$\begin{aligned} \dot{x} &= \sum_{i=1}^m g_i(x) v_i \\ \dot{v}_i &= u_i \quad i = 1, \dots, m \end{aligned}$$

is also controllable (and vice versa)

- in the linear case  $\dot{x} = Ax + \sum_{j=1}^m b_j u_j = Ax + Bu$ , all controllability definitions are equivalent and the associated tests reduce to the well-known Kalman rank condition:

$$\text{rank} (B \quad AB \quad A^2B \quad \dots \quad A^{n-1}B) = n$$

- a controllability test is a nonholonomy test!

a set of  $k$  Pfaffian constraints  $A(q)\dot{q} = 0$  is nonholonomic if and only if the associated kinematic model

$$\dot{q} = G(q)u = \sum_{i=1}^m g_i(q)u_i \quad m = n - k$$

is controllable, that is

$$\dim \bar{\mathcal{C}} = n$$

being  $\bar{\mathcal{C}}$  the involutive closure of the distribution associated with  $g_1, \dots, g_m$



for a nonholonomic system, it is always possible to design **open-loop** commands that drive the system from any state to any other state (**nonholonomic path planning**)

## Stabilizability of Nonholonomic Systems

for a nonlinear control system

$$\dot{x} = f(x) + \sum_{j=1}^m g_j(x)u_j = f(x) + g(x)u$$

one would like to build a **feedback control** law of the form

$$u = \alpha(x) + \beta(x)v$$

in such a way that either

- a) a desired closed-loop equilibrium point  $x_e$  is made asymptotically stable, or
  - b) a desired feasible closed-loop trajectory  $x_d(t)$  is made asymptotically stable
- feedback laws are essential in motion control to counteract the presence of disturbances as well as modeling inaccuracies
  - in linear systems, controllability directly implies asymptotic (actually, exponential) stabilizability at  $x_e$  by **smooth** (actually, linear) state feedback

$$\alpha(x) = K(x - x_e)$$

- if the linear approximation of the system at  $x_e$

$$\dot{\delta x} = A\delta x + B\delta u \quad \delta x = x - x_e, \delta u = K\delta x$$

is controllable, then the original system can be locally smoothly stabilized at  $x_e$  (a **sufficient** condition)

- in the presence of **uncontrollable eigenvalues at zero**, nothing can be concluded (except that smooth exponential stability is not achievable)
- for kinematic models of nonholonomic systems  $\dot{q} = G(q)u$ , the linear approximation around  $x_e$  has **always** uncontrollable eigenvalues at zero since

$$A \equiv 0 \quad \text{and} \quad \text{rank } B = \text{rank } G(q_e) = m < n$$

- however, there are **necessary** conditions for the existence of a  $C^0$ -stabilizing state feedback law (next slide)
- whenever these conditions fail, two alternatives are left:
  - a) **discontinuous feedback**  $u = \alpha(x), \alpha \in \bar{C}^0$
  - b) **time-varying feedback**  $u = \alpha(x, t), \alpha \in C^1$

## Brockett stabilization theorem (1983)

if the system

$$\dot{x} = f(x, u)$$

is locally asymptotically  $C^1$ -stabilizable at  $x_e$ , then the image of the map

$$f : \mathcal{M} \times \mathcal{U} \rightarrow \mathbb{R}^n$$

contains some **neighborhood** of  $x_e$  (a **necessary** condition)

a special case: the **driftless** system

$$\dot{x} = \sum_{i=1}^m g_i(x) u_i$$

with linearly independent vectors  $g_i(x_e)$ , i.e.,

$$\text{rank} ( g_1(x_e) \quad g_2(x_e) \quad \dots \quad g_m(x_e) ) = m$$

is locally asymptotically  $C^1$ -stabilizable at  $x_e$  **if and only if**  $m \geq n$



**nonholonomic mechanical systems**  
(either in kinematic or dynamic form)  
**cannot be stabilized at a point by smooth feedback**

## Classification of Nonholonomic Distributions

- the equivalence between a set of Pfaffian constraints

$$a_i^T(q)\dot{q} = 0 \quad i = 1, \dots, k$$

and the associated kinematic model

$$\dot{q} = \sum_{j=1}^m g_j(q)u_j \quad m = n - k$$

i.e., in matrix format

$$A^T(q)\dot{q} = 0 \quad \iff \quad \dot{q} = G(q)u$$

in the light of controllability (LA) conditions gives

$\dim \bar{\mathcal{C}} = n$	$\iff$	completely nonholonomic constraints (distribution)
$m < \dim \bar{\mathcal{C}} < n$	$\iff$	partially nonholonomic constraints (distribution)
$\dim \bar{\mathcal{C}} = m$	$\iff$	holonomic constraints (distribution)

- Frobenius Theorem  $\Rightarrow$

if  $\bar{\mathcal{C}}$  is regular of dimension  $n - p$ , there exist  $p$  functions  $h_j$  such that

$$h_j(q) = c_j \quad (j = 1, \dots, p) \iff a_i^T(q)\dot{q} = 0 \quad (i = 1, \dots, k)$$

- one may show that the **complexity** of the path planning problem is related to the level of Lie bracketing needed to span  $\mathbb{R}^n$



classify nonholonomic systems accordingly

- the **filtration**  $\{\mathcal{C}_i\}$  generated by the distribution  $\mathcal{C} = \text{span}\{g_1, \dots, g_m\}$  is defined as

$$\begin{aligned} \mathcal{C}_1 &= \mathcal{C} \\ \mathcal{C}_i &= \mathcal{C}_{i-1} + [\mathcal{C}_1, \mathcal{C}_{i-1}] \quad i > 2 \end{aligned}$$

where

$$[\mathcal{C}_1, \mathcal{C}_{i-1}] = \text{span}\{[g_j, v] : g_j \in \mathcal{C}_1, v \in \mathcal{C}_{i-1}\}$$

- a filtration is **regular** in a neighborhood  $V(q_0)$  if  $\dim \mathcal{C}_i(q) = \dim \mathcal{C}_i(q_0)$ ,  $\forall q \in V(q_0)$
- if  $\{\mathcal{C}_i\}$  is regular, the **degree of nonholonomy** of  $\mathcal{C}$  is the smallest integer  $\kappa$  such that

$$\dim \mathcal{C}_{\kappa+1} = \dim \mathcal{C}_\kappa$$

- $\Rightarrow$  nonholonomy conditions in terms of  $\kappa$ : a set of  $k$  Pfaffian constraints is
  1. completely nonholonomic if  $\dim \mathcal{C}_\kappa = n$
  2. partially nonholonomic if  $m < \dim \mathcal{C}_\kappa < n$
  3. holonomic if  $\dim \mathcal{C}_\kappa = m$  ( $m = n - k$ )

## Examples of Classification

- **unicycle kinematics** ( $n = 3$ )

$$g_1 = \begin{pmatrix} \cos \theta \\ \sin \theta \\ 0 \end{pmatrix} \quad g_2 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \quad g_3 = [g_1, g_2] = \begin{pmatrix} -\sin \theta \\ \cos \theta \\ 0 \end{pmatrix}$$

degree of nonholonomy  $\kappa = 2$ ,  $\dim \bar{\mathcal{C}} = 3$  for all  $q$

- **unicycle dynamics** ( $n = 5$ )

$$f = \begin{pmatrix} \cos \theta v_1 \\ \sin \theta v_1 \\ v_2 \\ 0 \\ 0 \end{pmatrix} \quad g_1 = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1/m \\ 0 \end{pmatrix} \quad g_2 = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1/I \end{pmatrix}$$

$$[g_1, f] = \begin{pmatrix} \cos \theta/m \\ \sin \theta/m \\ 0 \\ 0 \\ 0 \end{pmatrix} \quad [g_2, f] = \begin{pmatrix} 0 \\ 0 \\ 1/I \\ 0 \\ 0 \end{pmatrix} \quad [g_2, [f, [g_1, f]]] = \begin{pmatrix} -\sin \theta/mI \\ \cos \theta/mI \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

degree of nonholonomy  $\kappa = 3$ ; satisfies both the LA and STLC conditions since

$$g_1 \quad g_2 \quad [g_1, f] \quad [g_2, f] \quad [g_2, [f, [g_1, f]]]$$

span  $\mathbb{R}^5$ , and the sequence is 'good' [Sussmann]

- **car-like robot (RD)** ( $n = 4$ )

$$g_1 = \begin{pmatrix} \cos \theta \\ \sin \theta \\ \tan \phi / \ell \\ 0 \end{pmatrix} \quad g_2 = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

$$g_3 = [g_1, g_2] = \begin{pmatrix} 0 \\ 0 \\ -1/\ell \cos^2 \phi \\ 0 \end{pmatrix}$$

$$g_4 = [g_1, g_3] = \begin{pmatrix} -\sin \theta / \ell \cos^2 \phi \\ \cos \theta / \ell \cos^2 \phi \\ 0 \\ 0 \end{pmatrix}$$

degree of nonholonomy  $\kappa = 3$ ,  $\dim \bar{\mathcal{C}} = 4$  away from the singularity at  $\phi = \pm\pi/2$  of  $g_1$

- **car-like robot (FD)** ( $n = 4$ )

$$g_1 = \begin{pmatrix} \cos \theta \cos \phi \\ \sin \theta \cos \phi \\ \sin \phi / \ell \\ 0 \end{pmatrix} \quad g_2 = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

$$g_3 = [g_1, g_2] = \begin{pmatrix} \cos \theta \sin \phi \\ \sin \theta \sin \phi \\ -\cos \phi / \ell \\ 0 \end{pmatrix}$$

$$g_4 = [g_1, g_3] = \begin{pmatrix} -\sin \theta / \ell \\ \cos \theta / \ell \\ 0 \\ 0 \end{pmatrix}$$

degree of nonholonomy  $\kappa = 3$ ,  $\dim \bar{\mathcal{C}} = 4$  for all  $q$

- **$N$ -trailer system** ( $n = N + 4$ )

for a slightly modified version of this mobile robot the degree of nonholonomy is  $n$

- **all** the previous WMRs are STLC; **none** of these is smoothly stabilizable

- **3-body space robot** ( $n = 3$ )

$$g_1 = \begin{pmatrix} s_1(\phi) \\ 1 \\ 0 \end{pmatrix} \quad g_2 = \begin{pmatrix} s_2(\phi) \\ 0 \\ 1 \end{pmatrix}$$

$$g_3 = [g_1, g_2] = \begin{pmatrix} \frac{\partial s_2(\phi)}{\partial \phi_1} - \frac{\partial s_1(\phi)}{\partial \phi_2} \\ 0 \\ 0 \end{pmatrix}$$

but  $g_3 = 0$  for some combinations of  $\phi_1$  and  $\phi_2$

- the filtration is not regular: thus, the degree of nonholonomy is not well defined
- using higher order brackets,  $\dim \bar{\mathcal{C}} = 3$  for all  $q$  and the system is controllable

- **$N$ -body space robot dynamics** ( $n = 2N - 1$ )

the system satisfies the conditions for LA, STLC, but not the necessary condition for stabilizability via  $C^1$ -feedback

# NONHOLONOMIC MOTION PLANNING

- the objective is to build a sequence of **open-loop** input commands that steer the system from  $q_i$  to  $q_f$  satisfying the nonholonomic constraints
- the degree of nonholonomy gives a good measure of the complexity of the steering algorithm
- there exist **canonical** model structures for which the steering problem can be solved efficiently
  - chained form
  - power form
  - Caplygin form
- interest in the **transformation** of the original model equation into one of these forms
- such model structures allow also a simpler design of feedback stabilizers (necessarily, non-smooth or time-varying)
- we limit the analysis to the case of systems **with two inputs**, where the three above forms are equivalent (via a coordinate transformation)

## Chained Forms [Murray and Sastry 1993]

- a  $(2, n)$  **chained form** is a two-input driftless control system

$$\dot{z} = g_1(z)v_1 + g_2(z)v_2$$

in the following form

$$\begin{aligned}\dot{z}_1 &= v_1 \\ \dot{z}_2 &= v_2 \\ \dot{z}_3 &= z_2 v_1 \\ \dot{z}_4 &= z_3 v_1 \\ &\vdots \\ \dot{z}_n &= z_{n-1} v_1\end{aligned}$$

- denoting repeated Lie brackets as  $\text{ad}_{g_1}^k g_2$

$$\text{ad}_{g_1} g_2 = [g_1, g_2] \quad \text{ad}_{g_1}^k g_2 = [g_1, \text{ad}_{g_1}^{k-1} g_2]$$

one has

$$g_1 = \begin{pmatrix} 1 \\ 0 \\ z_2 \\ z_3 \\ \vdots \\ z_{n-1} \end{pmatrix} \quad g_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \Rightarrow \text{ad}_{g_1}^k g_2 = \begin{pmatrix} 0 \\ \vdots \\ (-1)^k \\ \vdots \\ 0 \end{pmatrix}$$

in which  $(-1)^k$  is the  $(k+2)$ -th entry

- a one-chain system is **completely nonholonomic (controllable)** since the  $n$  vectors

$$\{g_1, g_2, \dots, \text{ad}_{g_1}^i g_2, \dots\} \quad i = 1, \dots, n - 2$$

are independent

- its degree of nonholonomy is  $\kappa = n - 1$
- $v_1$  is called the **generating** input,  $z_1$  and  $z_2$  are called **base variables**
- if  $v_1$  is (piecewise) constant, the system in chained form behaves like a (piecewise) linear system
- chained systems are a generalization of first- and second-order controllable systems for which sinusoidal steering from  $z_i$  to  $z_f$  minimizes the integral norm of the input
- different input commands can be used, e.g.
  - sinusoidal inputs
  - piecewise constant inputs
  - polynomial inputs

## steering with sinusoidal inputs

- it is a two-phase method:

- I. steer the base variables  $z_1$  and  $z_2$  to their desired values  $z_{f1}$  and  $z_{f2}$  (in finite time)
- II. for each  $z_{k+2}$ ,  $k \geq 1$ , steer  $z_{k+2}$  to its final value  $z_{f,k+2}$  using

$$v_1 = \alpha \sin \omega t \quad v_2 = \beta \cos k\omega t$$

over one period  $T = 2\pi/\omega$ , where  $\alpha$ ,  $\beta$  are such that

$$\frac{\alpha^k \beta}{k!(2\omega)^k} = z_{f,k+2}(T) - z_{k+2}(0)$$

this guarantees  $z_i(T) = z_i(0) = z_{fi}$  for  $i < k$

in phase II, this step-by-step procedure adjusts one variable at a time by exploiting the closed-form integrability of the system equations under sinusoidal inputs

- phase II can be executed also all at once, choosing

$$\begin{aligned} v_1 &= a_0 + a_1 \sin \omega t \\ v_2 &= b_0 + b_1 \cos \omega t + \dots + b_{n-2} \cos(n-2)\omega t \end{aligned}$$

and solving numerically for the  $n+1$  unknowns in terms of the desired variation of the  $n-2$  states

## steering with piecewise constant inputs

- an idea coming from multirate digital control, with the total travel time  $T$  divided in subintervals of length  $\delta$  over which constant inputs are applied

$$\begin{aligned}v_1(\tau) &= v_{1,k} \\v_2(\tau) &= v_{2,k}\end{aligned}\quad \tau \in [(k-1)\delta, k\delta)$$

- it is convenient to keep  $v_1$  always constant and take  $n-1$  subintervals so that

$$T = (n-1)\delta \quad v_1 = \frac{z_{f1} - z_{01}}{T}$$

and the  $n-1$  constant values of input  $v_2$

$$v_{2,1}, v_{2,2}, \dots, v_{2,n-1}$$

are obtained solving a triangular linear system coming from the closed-form integration of the model equations

- if  $z_{f1} = z_{01}$ , an intermediate point must be added
- for small  $\delta$ , a fast motion but with large inputs

## steering with polynomial inputs

- idea similar to piecewise constant input, but with improved **smoothness** properties w.r.t. time (remember that kinematic models are controlled at the (pseudo)velocity level)
- the controls are chosen as

$$\begin{aligned}v_1 &= \text{sign}(z_{f1} - z_{01}) \\v_2 &= c_0 + c_1 t + \dots + c_{n-2} t^{n-2}\end{aligned}$$

with  $T = z_{f1} - z_{01}$  and  $c_0, \dots, c_n$  obtained solving the linear system coming from the closed-form integration of the model equations

$$M(T) \begin{pmatrix} c_0 \\ c_1 \\ \vdots \\ c_{n-2} \end{pmatrix} + m(z_i, T) = \begin{pmatrix} z_{f2} \\ z_{f3} \\ \vdots \\ z_{fn} \end{pmatrix}$$

with  $M(T)$  nonsingular for  $T \neq 0$

- if  $z_{f1} = z_{01}$ , an intermediate point must be added
- for small  $T$ , a fast motion but with large inputs

## transformation into chained form

- there exist necessary and sufficient conditions for transforming a control system

$$\dot{q} = g_1(q)u_1 + \dots + g_m(q)u_m \quad q \in \mathbb{R}^n$$

into chained form via input transformation and change of coordinates

$$v = \beta(q)u \quad z = T(q)$$

- for  $m = 2$ ,  $\mathcal{C} = \text{span}\{g_1, g_2\}$ , define the filtrations

$$\begin{aligned} E_1 &= \mathcal{C} & F_1 &= \mathcal{C} \\ E_2 &= E_1 + [E_1, E_1] & F_2 &= F_1 + [F_1, F_1] \\ &\vdots & &\vdots \\ E_{i+1} &= E_i + [E_i, E_i] & F_{i+1} &= F_i + [F_i, F_1] \end{aligned}$$

- the system can be transformed in chained form if and only if

$$\dim E_i = \dim F_i = i + 1 \quad i = 1, \dots, n - 1$$

**nonholonomic** systems up to order  $n = 4$  can be **always** be put in chained form!

- a simpler constructive sufficient condition: define the distributions

$$\Delta_0 = \text{span} \{g_1, g_2, \text{ad}_{g_1} g_2, \dots, \text{ad}_{g_1}^{n-2} g_2\}$$

$$\Delta_1 = \text{span} \{g_2, \text{ad}_{g_1} g_2, \dots, \text{ad}_{g_1}^{n-2} g_2\}$$

$$\Delta_2 = \text{span} \{g_2, \text{ad}_{g_1} g_2, \dots, \text{ad}_{g_1}^{n-3} g_2\}$$

if, for some open set, one has (i)  $\dim \Delta_0 = n$  (ii)  $\Delta_1, \Delta_2$  are involutive (iii) there exists a function  $h_1$  such that

$$dh_1 \cdot \Delta_1 = 0 \quad dh_1 \cdot g_1 = 1$$

then the system can be transformed into chained form

- the change of coordinates is given by

$$\begin{aligned} z_1 &= h_1 \\ z_2 &= L_{g_1}^{n-2} h_2 \\ &\vdots \\ z_{n-1} &= L_{g_1} h_2 \\ z_n &= h_2 \end{aligned}$$

with  $h_2$  independent from  $h_1$  and such that

$$dh_2 \cdot \Delta_2 = 0$$

- the input transformation is given by

$$\begin{aligned} v_1 &= u_1 \\ v_2 &= (L_{g_1}^{n-1} h_2) u_1 + (L_{g_2} L_{g_1}^{n-2} h_2) u_2 \end{aligned}$$

## WMRs in Chained Form

- **unicycle**

the change of coordinates

$$\begin{aligned}z_1 &= x \\z_2 &= \tan \theta \\z_3 &= y\end{aligned}$$

and input transformation

$$\begin{aligned}u_1 &= v_1 / \cos \theta \\u_2 &= v_2 \cos^2 \theta\end{aligned}$$

yield

$$\begin{aligned}\dot{z}_1 &= v_1 \\ \dot{z}_2 &= v_2 \\ \dot{z}_3 &= z_2 v_1\end{aligned}$$

other, globally defined transformations are possible

- **unicycle with  $N$  trailers**

the sufficient conditions are not satisfied but an 'ad hoc' transformation can be found (it starts using as  $(x, y)$  the position of the **last trailer** instead of the position of the trailing car)

- **car-like robot (RD)**

scaling first  $u_1$  by  $\cos \theta$

$$\begin{aligned}\dot{x} &= u_1 \\ \dot{y} &= u_1 \tan \theta \\ \dot{\theta} &= \frac{1}{l} u_1 \sec \theta \tan \phi \\ \dot{\phi} &= u_2\end{aligned}$$

then setting

$$\begin{aligned}z_1 &= x \\ z_2 &= \frac{1}{l} \sec^3 \theta \tan \phi \\ z_3 &= \tan \theta \\ z_4 &= y\end{aligned}$$

and

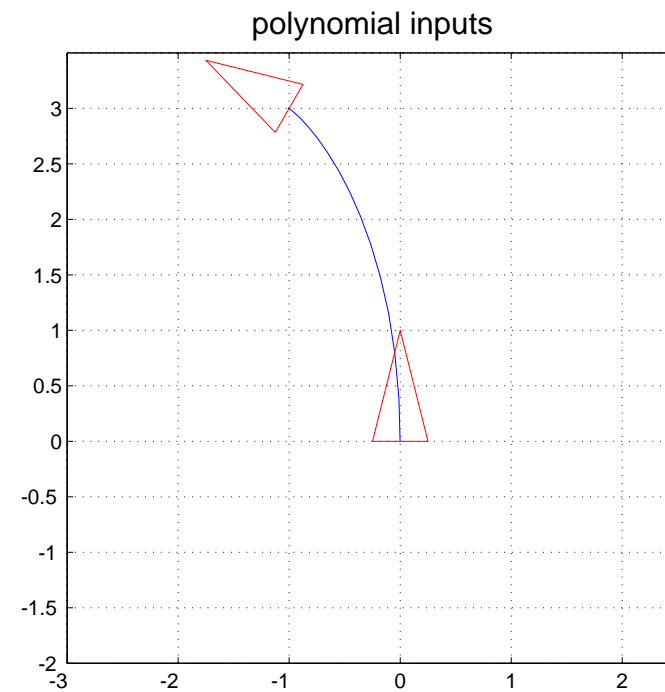
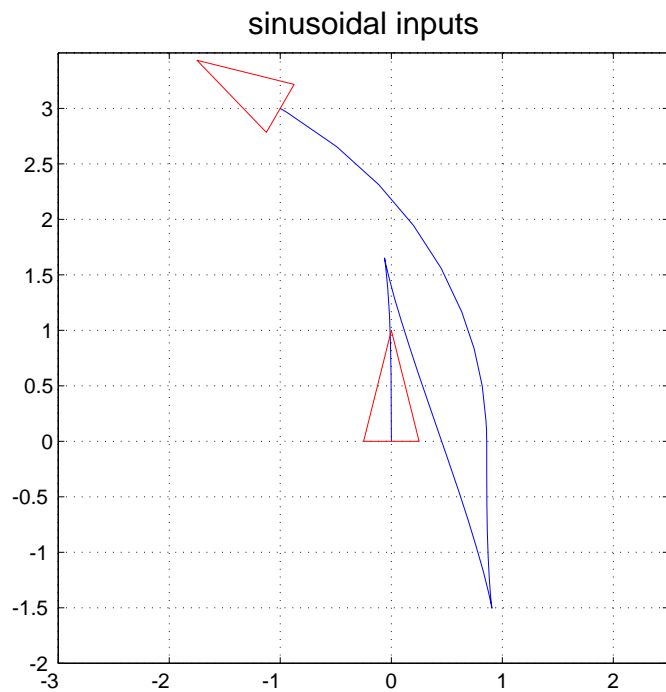
$$\begin{aligned}u_1 &= v_1 \\ u_2 &= -\frac{3}{l} v_1 \sec \theta \sin^2 \phi + \frac{1}{l} v_2 \cos^3 \theta \cos^2 \phi\end{aligned}$$

yields

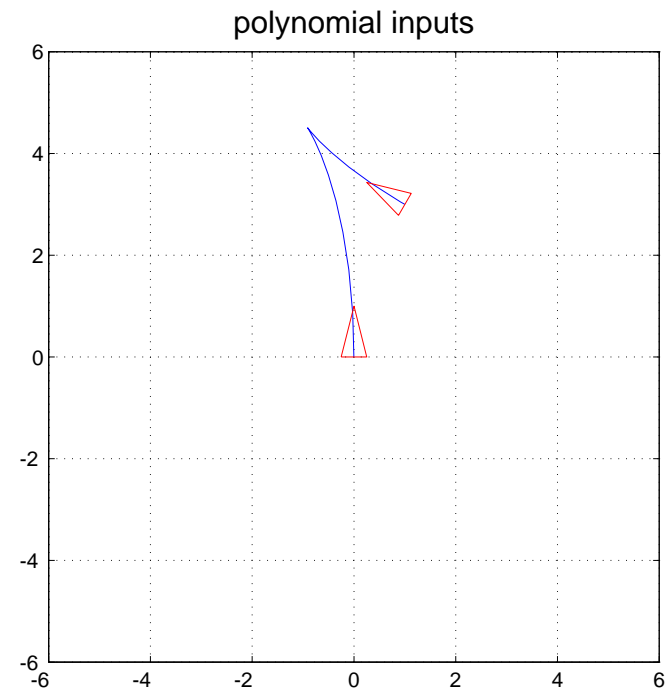
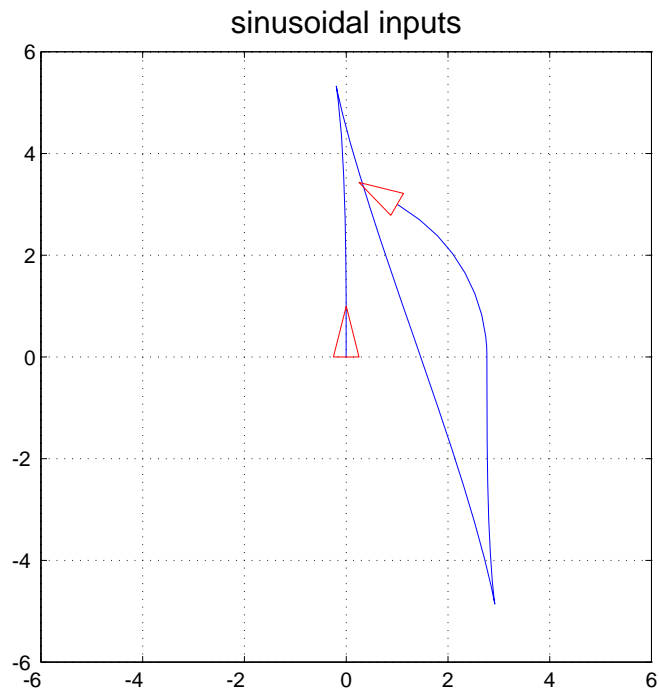
$$\begin{aligned}\dot{z}_1 &= v_1 \\ \dot{z}_2 &= v_2 \\ \dot{z}_3 &= z_2 v_1 \\ \dot{z}_4 &= z_3 v_1\end{aligned}$$

## Path Planning for the Unicycle

simulation 1:  $q_i = (-1, 3, 150^\circ)$ ,  $q_f = (0, 0, 90^\circ)$



simulation 2:  $q_i = (1, 3, 150^\circ)$ ,  $q_f = (0, 0, 90^\circ)$



## A General Viewpoint: Differential Flatness [Fliess *et al.* 1995]

- a nonlinear control system  $\dot{z} = f(z) + G(z)v$  is **differentially flat** if there exists a set of outputs  $y$  (**flat outputs**) such that the state and the input can be expressed **algebraically** in terms of  $y$  and a certain number  $r$  of its derivatives

$$\begin{aligned}z &= z(y, \dot{y}, \ddot{y}, \dots, y^{[r]}) \\v &= v(y, \dot{y}, \ddot{y}, \dots, y^{[r]})\end{aligned}$$

- for driftless systems, flatness is equivalent to chained-form transformability; the flat outputs of a chained form are  $z_1, z_n$  (i.e., the  $x, y$  coordinates of the robot for a WMR)
- for example, for the (2,3) chained form equivalent to a unicycle, the flat outputs are  $z_1, z_3$ ; one has

$$z_2 = \frac{\dot{z}_3}{\dot{z}_1} \quad \text{and} \quad v_1 = \dot{z}_1, \quad v_2 = \frac{\dot{z}_1 \ddot{z}_3 - \ddot{z}_1 \dot{z}_3}{\dot{z}_1^2}$$

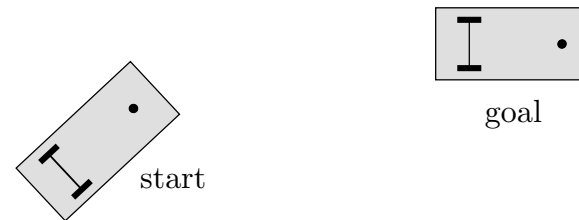
- for systems with drift, flatness is equivalent to dynamic feedback linearizability
- flatness is particularly useful for **path planning**: once the flat outputs are identified (a nontrivial task), any interpolation scheme can be used to join their initial and final values (with the appropriate boundary conditions); the evolution of the other variables as well as the control inputs are then computed through the algebraic transformations

# FEEDBACK CONTROL OF NONHOLONOMIC SYSTEMS

## Basic Problems

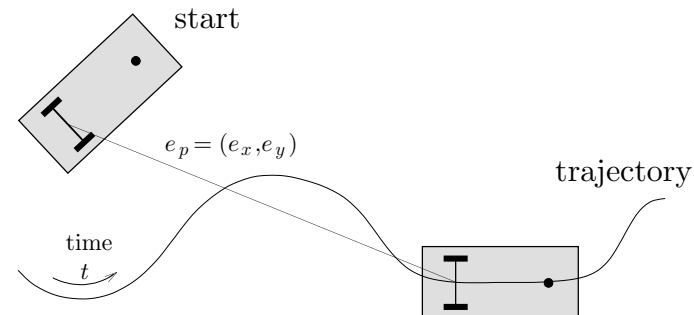
- target system: **unicycle**
  - the kinematic models of most single-body WMRs can be reduced to a unicycle
  - most of the presented design techniques can be systematically extended to chained-form transformable systems
- basic motion tasks

(a) point-to-point motion (PTPM)



(a)

(b) trajectory following (TF)



- PTPM via feedback: **posture stabilization**
  - w.l.o.g., the origin  $(0, 0, 0)$  is assumed to be the desired posture
  - a **nonsquare** ( $q \in \mathbb{R}^3, u \in \mathbb{R}^2$ ) state regulation problem
  - need to use discontinuous/time-varying feedback in view of Brockett Theorem
  - poor, erratic transient performance is often obtained (inefficient, unsafe in the presence of obstacles)
- TF via feedback: **asymptotic tracking**
  - the desired trajectory  $q_d(t)$  must be feasible, i.e., comply with the nonholonomic constraints
  - a **square** ( $e_p \in \mathbb{R}^2, u \in \mathbb{R}^2$ ) error zeroing problem
  - in this case, smooth feedback can be used because the linear approximation along a nonvanishing trajectory is controllable (see later)



**asymptotic tracking is easier** (and more useful) **than posture stabilization for nonholonomic systems**

## Asymptotic Tracking

- a reference output trajectory  $(x_d(t), y_d(t))$  is given
- control action: **feedforward** + **error feedback**  
error may be defined w.r.t. the reference output (**output error**) or the associated reference state (**state error**)
- given an initial posture and a desired trajectory  $(x_d(t), y_d(t))$  there is a **unique** associated state trajectory  $q_d(t) = (x_d(t), y_d(t), \theta_d(t))$  which can be computed in a purely algebraic way as

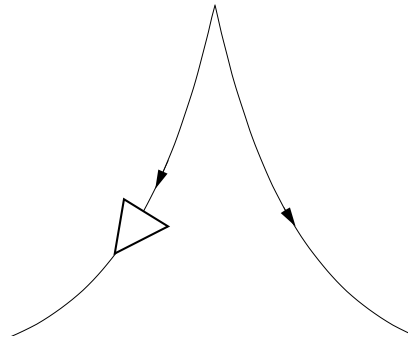
$$\theta_d(t) = \text{ATAN2}(\dot{y}_d(t), \dot{x}_d(t)) + k\pi \quad k = 0, 1$$

this is due to the fact that  $(x, y)$  is a **flat** output for the unicycle

- **feedforward command generation**: being  $\theta = \text{ATAN2}(\dot{y}, \dot{x}) + k\pi$ ,  $k = 0, 1$ , we get

$$u_{d1}(t) = \pm \sqrt{\dot{x}_d^2(t) + \dot{y}_d^2(t)}$$
$$u_{d2}(t) = \frac{\ddot{y}_d(t)\dot{x}_d(t) - \ddot{x}_d(t)\dot{y}_d(t)}{\dot{x}_d^2(t) + \dot{y}_d^2(t)}$$

- the choice of sign for  $u_{d1}(t)$  produces forward or backward motion
- to be exactly reproducible,  $(x_d(t), y_d(t))$  should be twice differentiable
- $\theta_d(t)$  may be computed off-line and used in order to define a state error
- if  $u_{d1}(\bar{t}) = 0$  for some  $\bar{t}$  (e.g., at a cusp)



neither  $u_{d2}(\bar{t})$  nor  $\theta_d(\bar{t})$  are defined

$\Rightarrow$  a continuous motion is guaranteed by keeping the same orientation attained at  $\bar{t}^-$

## asymptotic tracking: controllability

linear approximation along  $q_d(t) = (x_d(t), y_d(t), \theta_d(t))$

- define:

$u_{d1}, u_{d2}$  the inputs associated to  $q_d(t)$

$\tilde{q} = q - q_d$  the state tracking error

$\tilde{u}_1 = u_1 - u_{d1}$  and  $\tilde{u}_2 = u_2 - u_{d2}$  the input variations

- the linear approximation along  $q_d(t)$  is

$$\dot{\tilde{q}} = \begin{pmatrix} 0 & 0 & -u_{d1} \sin \theta_d \\ 0 & 0 & u_{d1} \cos \theta_d \\ 0 & 0 & 0 \end{pmatrix} \tilde{q} + \begin{pmatrix} \cos \theta_d & 0 \\ \sin \theta_d & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \tilde{u}_1 \\ \tilde{u}_2 \end{pmatrix}$$

a time-varying system

⇒ the N&S controllability condition is that the controllability Gramian is nonsingular

- a simpler analysis can be performed by ‘rotating’ the state tracking error

$$\tilde{q}_R = \begin{pmatrix} \cos \theta_d & \sin \theta_d & 0 \\ -\sin \theta_d & \cos \theta_d & 0 \\ 0 & 0 & 1 \end{pmatrix} \tilde{q}$$

according to the reference orientation  $\theta_d$

- we get

$$\dot{\tilde{q}}_R = \begin{pmatrix} 0 & u_{d2} & 0 \\ -u_{d2} & 0 & u_{d1} \\ 0 & 0 & 0 \end{pmatrix} \tilde{q}_R + \begin{pmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \tilde{u}_1 \\ \tilde{u}_2 \end{pmatrix}$$

- when the inputs  $u_{d1}$  and  $u_{d2}$  are constant, the linearization becomes time-invariant and controllable, since

$$(B \ AB \ A^2B) = \begin{pmatrix} 1 & 0 & 0 & 0 & -u_{d2}^2 & u_{d1}u_{d2} \\ 0 & 0 & -u_{d2} & u_{d1} & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \end{pmatrix}$$

has rank 3 provided that either  $u_{d1}$  or  $u_{d2}$  is nonzero

⇒ the kinematic model of the unicycle can be locally asymptotically stabilized by linear feedback along trajectories consisting of **linear or circular paths** executed at a constant velocity

(actually: the same can be proven for **any** nonvanishing trajectory)

## linear control design [Samson 1992]

- designed using a (slightly different) linear approximation along the reference trajectory
- define the state tracking error  $e$  as

$$\begin{pmatrix} e_1 \\ e_2 \\ e_3 \end{pmatrix} = \begin{pmatrix} \cos \theta & \sin \theta & 0 \\ -\sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_d - x \\ y_d - y \\ \theta_d - \theta \end{pmatrix}$$

- use a nonlinear transformation of velocity inputs

$$\begin{aligned} u_1 &= u_{d1} \cos e_3 - v_1 \\ u_2 &= u_{d2} - v_2 \end{aligned}$$

- the error dynamics becomes

$$\dot{e} = \begin{pmatrix} 0 & u_{d2} & 0 \\ -u_{d2} & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} e + \begin{pmatrix} 0 \\ \sin e_3 \\ 0 \end{pmatrix} u_{d1} + \begin{pmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$$

- linearizing around the reference trajectory, one obtains the same linear time-varying equations as before, now with state  $e$  and input  $(v_1, v_2)$

- define the ‘linear’ feedback law

$$\begin{aligned} v_1 &= -k_1 e_1 \\ v_2 &= -k_2 \operatorname{sign}(u_{d1}(t)) e_2 - k_3 e_3 \end{aligned}$$

with gains

$$k_1 = k_3 = 2\zeta a \quad k_2 = \frac{a^2 - u_{d2}(t)^2}{|u_{d1}(t)|}$$

- the closed-loop characteristic polynomial is  $(\lambda + 2\zeta a)(\lambda^2 + 2\zeta a\lambda + a^2)$ ,  $\zeta \in (0, 1)$   $a > 0$
- a convenient **gain scheduling** is achieved letting

$$a = a(t) = \sqrt{u_{d2}^2(t) + bu_{d1}^2(t)} \quad \implies \quad k_1 = k_3 = 2\zeta \sqrt{u_{d2}^2(t) + bu_{d1}^2(t)}, \quad k_2 = b|u_{d1}(t)|$$

these gains go to zero when the state trajectory stops (and local controllability is lost)

- the actual controls are **nonlinear** and **time-varying**
- even if the eigenvalues are constant, local asymptotic stability is not guaranteed as the system is still time-varying

$\Rightarrow$  a Lyapunov-based analysis is needed

## nonlinear control design [Samson 1993]

for the previous error dynamics, define

$$\begin{aligned}v_1 &= -k_1(u_{d1}(t), u_{d2}(t)) e_1 \\v_2 &= -\bar{k}_2 u_{d1}(t) \frac{\sin e_3}{e_3} e_2 - k_3(u_{d1}(t), u_{d2}(t)) e_3\end{aligned}$$

with constant  $\bar{k}_2 > 0$  and positive, continuous gain functions  $k_1(\cdot, \cdot)$  and  $k_3(\cdot, \cdot)$

**theorem** *if  $u_{d1}$ ,  $u_{d2}$ ,  $\dot{u}_{d1}$   $\dot{u}_{d2}$  are bounded, and if  $u_{d1}(t) \not\rightarrow 0$  or  $u_{d2}(t) \not\rightarrow 0$  as  $t \rightarrow \infty$ , the above control globally asymptotically stabilizes the origin  $e = 0$*

**proof** based on the Lyapunov function

$$V = \frac{\bar{k}_2}{2} (e_1^2 + e_2^2) + \frac{e_3^2}{2}$$

nonincreasing along the closed-loop solutions

$$\dot{V} = -k_1 \bar{k}_2 e_1^2 - k_3 e_3^2 \leq 0$$

$\Rightarrow \|e(t)\|$  is bounded,  $\dot{V}(t)$  is uniformly continuous, and  $V(t)$  tends to some limit value

$\Rightarrow$  using Barbalat lemma,  $\dot{V}(t)$  tends to zero

$\Rightarrow$  analyzing the system equations, one can show that  $(u_{d1}^2 + u_{d2}^2)e_i^2$  ( $i = 1, 2, 3$ ) tends to zero so that, from the persistency of the trajectory, the thesis follows ■

## dynamic feedback linearization [Oriolo *et al.*, 2002]

- define the output as  $\eta = (x, y)$ ; differentiation w.r.t. time yields

$$\dot{\eta} = \begin{pmatrix} \dot{x} \\ \dot{y} \end{pmatrix} = \begin{pmatrix} \cos \theta & 0 \\ \sin \theta & 0 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \end{pmatrix}$$

$\Rightarrow$  cannot recover  $u_2$  from first-order differential information

- add an integrator on the linear velocity input

$$u_1 = \xi, \quad \dot{\xi} = a \quad \Rightarrow \quad \dot{\eta} = \xi \begin{pmatrix} \cos \theta \\ \sin \theta \end{pmatrix}$$

new input  $a$  is the unicycle **linear acceleration**

- differentiating further

$$\ddot{\eta} = \begin{pmatrix} \cos \theta & -\xi \sin \theta \\ \sin \theta & \xi \cos \theta \end{pmatrix} \begin{pmatrix} a \\ u_2 \end{pmatrix}$$

- **assuming**  $\xi \neq 0$ , we can let

$$\begin{pmatrix} a \\ u_2 \end{pmatrix} = \begin{pmatrix} \cos \theta & -\xi \sin \theta \\ \sin \theta & \xi \cos \theta \end{pmatrix}^{-1} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$$

obtaining

$$\ddot{\eta} = \begin{pmatrix} \ddot{\eta}_1 \\ \ddot{\eta}_2 \end{pmatrix} = \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$$

- the resulting dynamic compensator is

$$\begin{aligned}\dot{\xi} &= v_1 \cos \theta + v_2 \sin \theta \\ u_1 &= \xi \\ u_2 &= \frac{v_2 \cos \theta - v_1 \sin \theta}{\xi}\end{aligned}$$

- as the dynamic compensator is 1-dim, we have  $n + 1 = 4$ , equal to the total number of output differentiations

⇒ in the new coordinates

$$\begin{aligned}z_1 &= x \\ z_2 &= y \\ z_3 &= \dot{x} = \xi \cos \theta \\ z_4 &= \dot{y} = \xi \sin \theta\end{aligned}$$

the system is fully linearized and described by two chains of second-order input-output integrators

$$\begin{aligned}\ddot{z}_1 &= v_1 \\ \ddot{z}_2 &= v_2\end{aligned}$$

- the dynamic feedback linearizing controller has a potential singularity at  $\xi = u_1 = 0$ , i.e., when the unicycle is not rolling

a singularity in the dynamic extension process is **structural** for nonholonomic systems

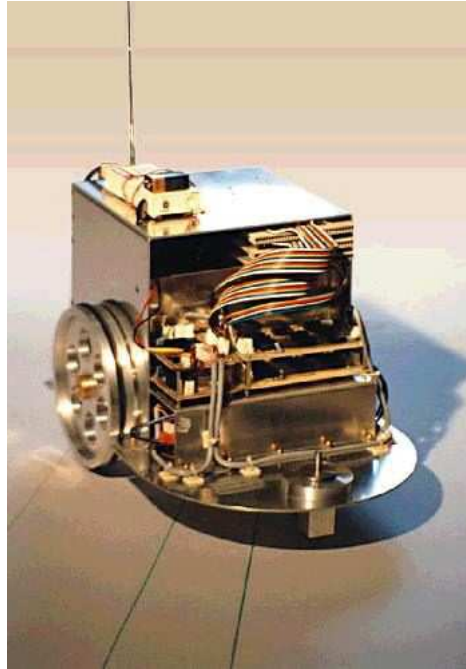
- for the (exactly) linearized system, a globally exponentially stabilizing feedback is

$$\begin{aligned} v_1 &= \ddot{x}_d(t) + k_{p1}(x_d(t) - x) + k_{d1}(\dot{x}_d(t) - \dot{x}) \\ v_2 &= \ddot{y}_d(t) + k_{p2}(y_d(t) - y) + k_{d2}(\dot{y}_d(t) - \dot{y}) \end{aligned}$$

with PD gains  $k_{pi} > 0$ ,  $k_{di} > 0$ , for  $i = 1, 2$

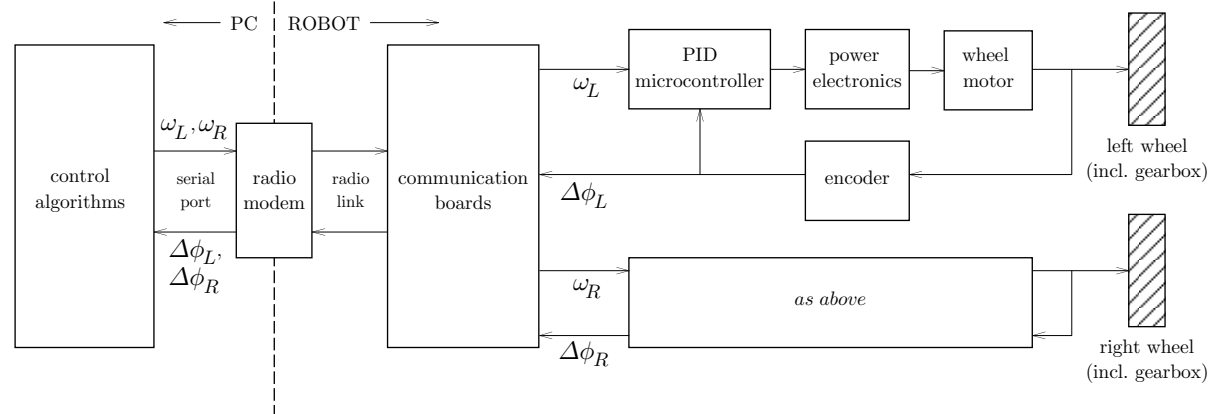
- the desired trajectory  $(x_d(t), y_d(t))$  must be smooth and **persistent**, i.e.,  $u_{d1}^2 = \dot{x}_d^2 + \dot{y}_d^2$  must never go to zero
- cartesian transients are linear
- $\dot{x}$  and  $\dot{y}$  can be computed as a function  $\xi$  and  $\theta$ ; alternatively, one can use estimates of  $\dot{x}$  and  $\dot{y}$  obtained from odometric measurements
- for **exact tracking**, one needs  $q(0) = q_d(0)$  and  $\xi(0) = u_{d1}(0)$  ( $\Rightarrow$  pure feedforward)

## experiments with SuperMARIO



- a two-wheel differentially-driven vehicle (with castor)
- the aluminum chassis measures  $46 \times 32 \times 30.5$  cm (l/w/h) and contains two motors, transmission elements, electronics, and four 12 V batteries; total weight about 20 kg
- each wheel independently driven by a DC motor (peak torque  $\approx 0.56$  Nm); each motor equipped with an encoder (200 pulse/turn) and a gearbox (reduction ratio 20)
- typical nonidealities of electromechanical systems: friction, gear backlash, wheel slippage, actuator deadzone and saturation
- due to robot and motor dynamics, discontinuous velocity commands cannot be realized

## two-level control architecture



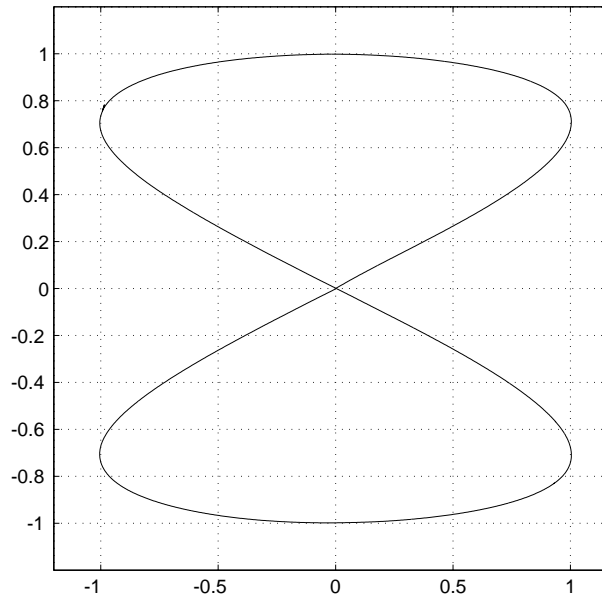
- control algorithms (with reference generation) are written in C++ and run with a sampling time of  $T_s = 50$  ms on a remote server
- the PC communicates through a radio modem with the serial communication boards on the robot
- actual commands are the angular velocities  $\omega_R$  and  $\omega_L$  of right and left wheel (instead of driving and steering velocities  $u_1$  and  $u_2$ ):

$$u_1 = \frac{r(\omega_R + \omega_L)}{2} \quad u_2 = \frac{r(\omega_R - \omega_L)}{d}$$

with  $d =$  axle length,  $r =$  wheel radius

- reconstruction of the current robot state based on encoder data (**dead reckoning**)

experiments on an eight-shaped trajectory



- the reference trajectory

$$x_d(t) = \sin \frac{t}{10} \quad y_d(t) = \sin \frac{t}{20} \quad t \in [0, T]$$

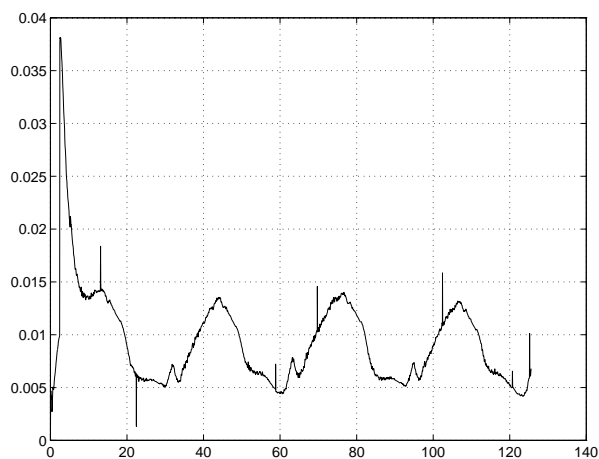
starts from the origin with  $\theta_d(0) = \pi/6$  rad

- a full cycle is completed in  $T = 2\pi \cdot 20 \approx 125$  s
- the reference initial velocities are

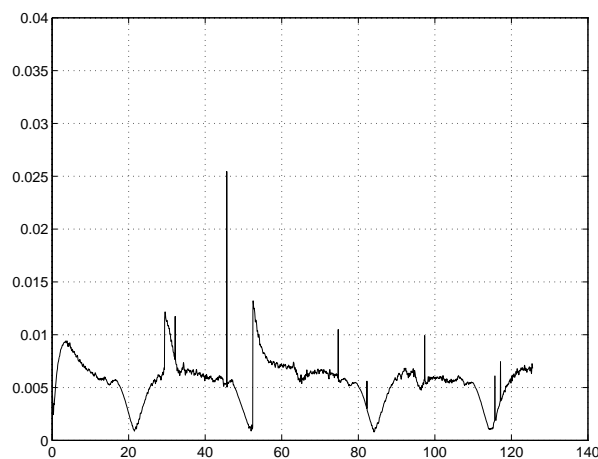
$$u_{d1}(0) \simeq 0.1118 \text{ m/s}, \quad u_{d2}(0) = 0 \text{ rad/s.}$$

experiment 1: the robot initial state is **on** the reference trajectory

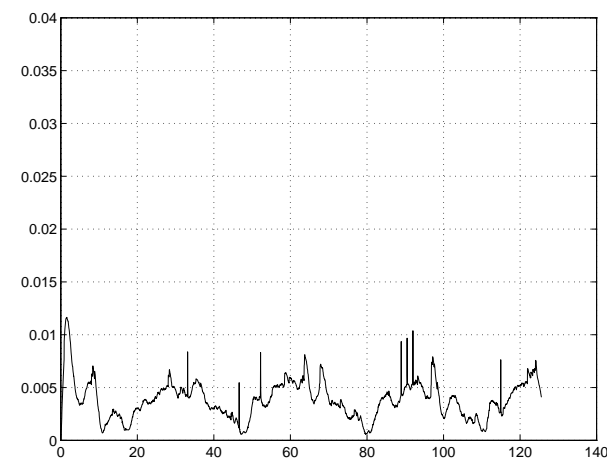
tracking error norm



linear design

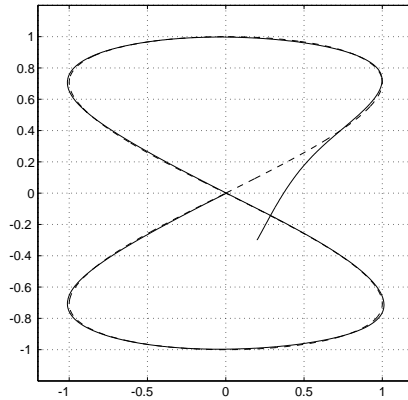


nonlinear design

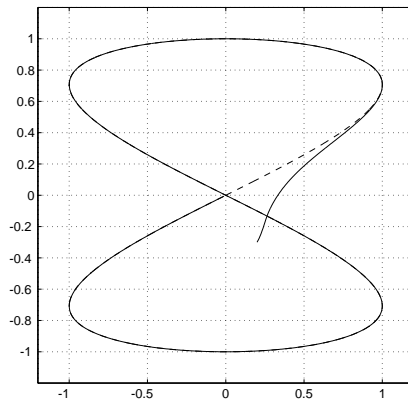
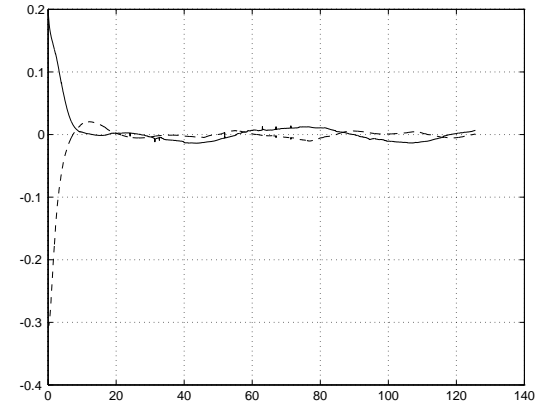


feedback linearization

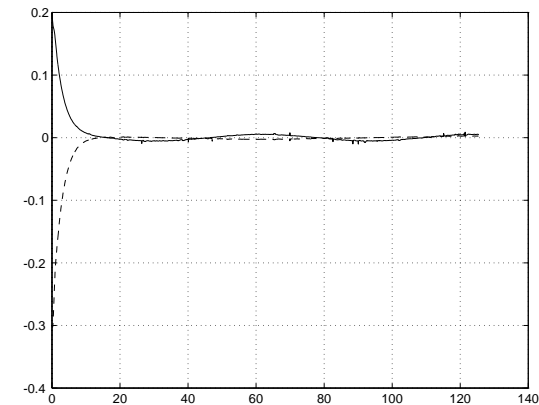
experiment 2: the robot initial state is **off** the reference trajectory



linear design



feedback linearization



## Posture Stabilization: A Bird's Eye View

- the main obstruction is the non-smooth stabilizability of WMRs at a point
- two main approaches
  - **time-varying** stabilizers: an exogenous time-varying signal is injected in the controller [Samson 1991]
  - **discontinuous** stabilizers: the controller is time invariant but discontinuous at the origin [Sørdalen 1993]
- drawbacks: slow convergence (time-varying), oscillatory transient (both)
- improvements
  - **mixed time-varying/discontinuous** stabilizers [Pomet and Samson 1993; Murray and M'Closkey 1995, Lucibello and Oriolo 2001]
  - **non-Lyapunov, discontinuous** stabilizers: through coordinate transformations that circumvent Brockett's obstruction [Aicardi *et al.* 1995; Astolfi 1995] or via dynamic feedback linearization [Oriolo *et al.* 2002]
    - ↪ excellent transient performance!

# OPTIMAL TRAJECTORIES FOR WMRS

(by M. Vendittelli)

- the **main objective** is to determine an optimal control law steering the kinematic model of the nonholonomic system between any two points of the configuration space
- a **first step** is to obtain a family of trajectories containing an optimal solution to the steering problem
- Pontryagin's Maximum Principle (PMP) can be used to this end providing necessary conditions for trajectories to be optimal
- characterization of optimal trajectories is not easy essentially due to the local nature of PMP
- local information needs to be completed by global study based on geometric reasoning

## Minimum-Time Problems

- **objective:** compute the control law (if it exists) that steers the nonholonomic system

$$\dot{q} = G(q)u, \quad q \in \mathcal{M} \simeq \mathbb{R}^n, \quad u \in U \subset \mathbb{R}^m$$

from  $q_i$  to  $q_f$  minimizing the functional

$$J = \int_{t_i}^{t_f} dt$$

- **theorem** (existence of optimal trajectories)

under the usual assumptions for existence and uniqueness of solution of an ordinary differential equation and the additional hypothesis

$$U \text{ **compact convex** subset of } \mathbb{R}^m \quad (\Delta)$$

any two points  $q_i, q_f \in \mathcal{M}$  that can be joined by an admissible trajectory can be joined by a time-optimal trajectory

- consider the **Hamiltonian**

$$H(\psi, q, u) = \langle \psi, G(q)u \rangle$$

where  $\psi \in \mathbb{R}^n$  and  $\langle \cdot, \cdot \rangle$  is the inner product in  $\mathbb{R}^n$

- if  $u(t) : [t_i, t_f] \rightarrow U$  is an admissible control law and  $q(t) : [t_i, t_f] \rightarrow \mathcal{M}$  the corresponding trajectory,  
a vector function  $\psi : [t_i, t_f] \rightarrow \mathbb{R}^n$  is an **adjoint vector** for  $(q, u)$  if it satisfies

$$\dot{\psi}(t)^T = -\frac{\partial H}{\partial q}(\psi(t), q(t), u(t)) \quad \forall t \in [t_i, t_f]$$

- note that

either  $\psi(t) \neq 0 \quad \forall t \in [t_i, t_f]$  (*nontrivial*  $\psi$ )

or  $\psi(t) \equiv 0 \quad \forall t \in [t_i, t_f]$  (*trivial*  $\psi$ )

due to the linearity of  $H$  ( $\Rightarrow$  of  $\dot{\psi}$ ) w.r.t.  $\psi$

## PMP for time-optimal control

consider an admissible control law  $u(t)$  and the corresponding trajectory  $q(t)$ ; a necessary condition for  $q(t)$  to be time-optimal is that there exist a nontrivial adjoint vector  $\psi(t)$  and a constant  $\psi_0 \leq 0$  s.t.

$$H(\psi(t), q(t), u(t)) = \max_{v \in U} \{H(\psi(t), q(t), v)\} = -\psi_0 \quad (*)$$

$$\forall t \in [t_i, t_f]$$

- a control law  $u(t)$  satisfying condition (\*) is called an **extremal control law**
- denoting by  $q, \psi$  the trajectory and the adjoint vector corresponding to the extremal control law  $u$ , the triple  $(q, u, \psi)$  is called **extremal**
- an extremal triple  $(q, u, \psi)$  s.t.  $\psi_0 = 0$  is called **abnormal**
- a control law  $u(t)$  is called **singular** if there exist a nonempty subset  $W \subset U$  and a nonempty interval  $I \subset [t_i, t_f]$  such that

$$H(\psi(t), q(t), u(t)) = H(\psi(t), q(t), w(t))$$

$$\forall t \in I, \forall w(t) \in W$$

## Application to WMRs

- target system: unicycle

$$\dot{q} = g_1(q)u_1 + g_2(q)u_2 \quad (u_1, u_2) \in U \subset \mathbb{R}^2$$

$$\text{with } q = (x, y, \theta) \quad g_1(q) = \begin{pmatrix} \cos \theta \\ \sin \theta \\ 0 \end{pmatrix} \quad g_2(q) = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

- new terminology based on **control domains**

$$U = [-k_1, k_1] \times [-k_2, k_2] \quad \textit{unicycle}$$

$$U = \{-k_1, k_1\} \times [-k_2, k_2] \quad \textit{Reeds and Shepp's car}$$

$$U = k_1 \times [-k_2, k_2] \quad \textit{Dubins' car}$$

$$\text{with } k_1, k_2 > 0$$

unicycle and Reeds and Shepp's car are STLC

Dubins' car is controllable but not STLC

- unicycle and Dubins' car verify the conditions for existence of optimal trajectories
- Reeds and Shepp's car does not verify condition ( $\Delta$ ); existence of optimal trajectories has been established as a byproduct of the analysis of the optimal control problem for the unicycle

## unicycle

- $(u_1, u_2) \in U = [-1, 1] \times [-1, 1]$  (w.l.o.g.)
- the corresponding Hamiltonian is

$$H = \psi_1 \cos \theta u_1 + \psi_2 \sin \theta u_1 + \psi_3 u_2$$

- it is convenient to define the **switching functions**

$$\phi_1 = \langle \psi, g_1 \rangle = \psi_1 \cos \theta + \psi_2 \sin \theta, \quad \phi_2 = \langle \psi, g_2 \rangle = \psi_3$$

and write the Hamiltonian as

$$H = \phi_1 u_1 + \phi_2 u_2$$

- the switching functions determine the sign changes of  $u_1, u_2$  (see later)
- applying PMP

$$-\psi_0 = H(\psi(t), q(t), u(t)) = \max_{v \in U} (H(\psi(t), q(t), v)) = \max_{(v_1, v_2) \in U} (\phi_1 v_1 + \phi_2 v_2) \quad (1)$$

where

$$\dot{\psi}(t) = -\frac{\partial H}{\partial q}(\psi(t), q(t), u(t)) = -\frac{\partial}{\partial q}(\phi_1 u_1 + \phi_2 u_2)$$

## extracting information from PMP

- maximization of the Hamiltonian (i.e. cond. (1)) implies that on extremal trajectories

$$u_1 = \text{sign}(\phi_1) \quad u_2 = \text{sign}(\phi_2) \quad (2)$$

where

$$\text{sign}(s) = \begin{cases} 1 & \text{if } s > 0 \\ -1 & \text{if } s < 0 \\ \text{any number in } [-1, 1] & \text{if } s = 0 \end{cases}$$

- on any subinterval of  $[t_i, t_f]$  where  $\phi_j \neq 0$  ( $j=\{1,2\}$ )  $u_j$  is **bang** (i.e. maximal or minimal)
- a necessary condition for  $t$  to be a switching time for  $u_j(t)$  is that  $\phi_j(t) = 0$
- if  $\phi_j(t) = 0$  on a nonempty interval  $I \subset [t_i, t_f]$  the corresponding control  $u_j(t)$  is singular on  $I$

- to characterize the structure of extremals

define

$$\phi_3 = \langle \psi, [g_2, g_1] \rangle$$

compute

$$\dot{\phi}_1 = u_2 \cdot \langle \psi, [g_2, g_1] \rangle = u_2 \phi_3$$

$$\dot{\phi}_2 = -u_1 \cdot \langle \psi, [g_2, g_1] \rangle = -u_1 \phi_3 \quad (3)$$

$$\dot{\phi}_3 = -u_2 \phi_1$$

- from (1), (2)

$$|\phi_1| + |\phi_2| + \psi_0 = 0 \quad (4)$$

- from  $\psi \neq 0$  + controllability

$$|\phi_1| + |\phi_2| + |\phi_3| \neq 0 \quad (5)$$

- (2), (3), (4), (5) are called **Switching Structure Equations**

**lemma 1** nontrivial abnormal extremals do not exist

proof: use (4), (5), (3)

**lemma 2** for a nontrivial optimal extremal,  $\phi_1$  and  $\phi_2$  cannot have a common zero

proof: use (4)

**lemma 3** along an extremal,  $\kappa = \phi_1^2 + \phi_3^2$  is constant and  $\kappa = 0 \iff \phi_1 \equiv 0$

proof: use lemma 2, (3)

**lemma 4** along an extremal, either all the zeros of  $\phi_1$  are isolated and s.t.  $\dot{\phi}_1$  exists and is  $\neq 0$  or  $\phi_1 \equiv 0$

proof: use lemma 3, lemma 2, (3)

↓

there exist two kinds of extremal trajectories

A trajectories with a finite number of switchings

B trajectories along which  $\phi_1 \equiv 0$  and either  $u_2 \equiv 1$  or  $u_2 \equiv -1$

to simplify the geometric description of the extremals it is useful to introduce the following

## notation

$C_a$  arc of circle of length  $a$

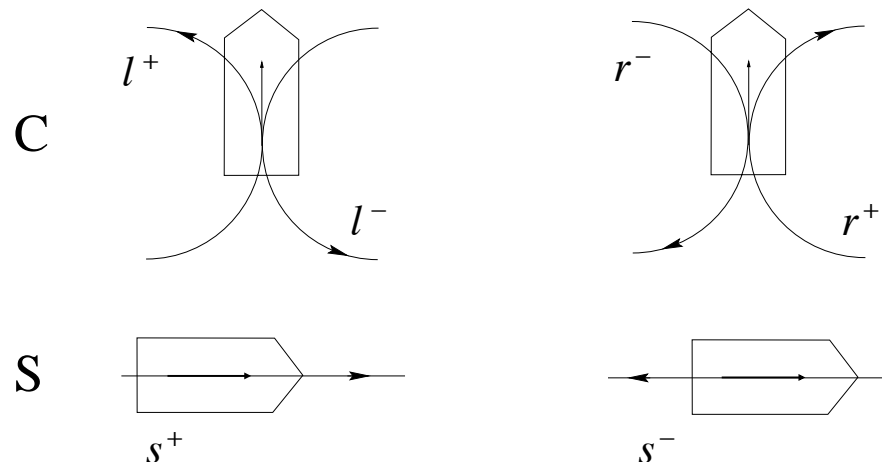
$S_a$  straight line segment of length  $a$

$C|C$  arcs of circle connected by a cusp

$l_a^{+(-)}$  forward (backward) left motion along the arc of length  $a$

$r_a^{+(-)}$  forward (backward) right motion along the arc of length  $a$

$s_a^{+(-)}$  forward (backward) motion along the straight line segment of length  $a$



## type A trajectories

the integration of the adjoint system

$$\begin{cases} \dot{\psi}_1 = -\frac{\partial H}{\partial x} = 0 \\ \dot{\psi}_2 = -\frac{\partial H}{\partial y} = 0 \\ \dot{\psi}_3 = -\frac{\partial H}{\partial \theta} = \psi_1 \sin \theta u_1 - \psi_2 \cos \theta u_1 = \psi_1 \dot{y} - \psi_2 \dot{x} \end{cases}$$

implies (w.l.o.g.  $x(t_i) = y(t_i) = 0$ )

- $\psi_1$  and  $\psi_2$  constant
- $\psi_3(t) = \psi_3(t_i) + \psi_1 y - \psi_2 x = \phi_2(t)$
- if  $\phi_1 = 0$  (switch of  $u_1$ ), (1) implies  $\phi_2 u_2 + \psi_0 = \psi_3 u_2 + \psi_0 = 0$

– if  $u_2 = 1$  the **cusp** point is on the line

$$\mathcal{D}^+ : \psi_1 y(t) - \psi_2 x(t) + \psi_3(t_i) + \psi_0 = 0$$

– if  $u_2 = -1$  the cusp point is on the line

$$\mathcal{D}^- : \psi_1 y(t) - \psi_2 x(t) + \psi_3(t_i) - \psi_0 = 0$$

- if  $\phi_2 = 0$  (switch of  $u_2$ ) the **inflection** point lies on the line

$$\mathcal{D}_0 : \psi_1 y(t) - \psi_2 x(t) + \psi_3(t_i) = 0$$

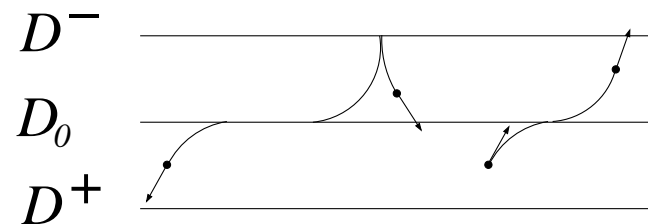
- if  $\phi_2(t)$  vanishes on a nonempty interval  $I \subset [t_i, t_f]$  from (1)

$$\psi_1 \cos(\theta(t)) + \psi_2 \sin(\theta(t)) + \psi_0 = 0$$

from lemma 2,  $\psi_1$  and  $\psi_2$  cannot be both zero then  $\theta$  must remain constant on  $I$

summarizing:

- type A trajectories are sequences of
  - arcs of circle ( $C$ ) of radius 1 corresponding to regular control laws ( $u_1 = \pm 1, u_2 = \pm 1$ )
  - straight segments ( $S$ ) corresponding to the singularity of  $u_2$  ( $u_1 = \pm 1, u_2 = 0$ )
- straight line segments and points of inflection are on  $\mathcal{D}_0$
- cusp tangents are perpendicular to  $\mathcal{D}^+$  and  $\mathcal{D}^-$
- **lemma** trajectories of type A and with no cusps are necessarily of one of the following kinds
  - $C_a$   $0 \leq a \leq \pi$
  - $C_a C_b$   $0 < a \leq \frac{\pi}{2}, 0 < b \leq \frac{\pi}{2}$
  - $C_a S_d C_b$   $d > 0, 0 < a \leq \frac{\pi}{2}, 0 < b \leq \frac{\pi}{2}$



- to refine the large family of trajectories implied by type A a global geometric study would be needed

- a *boundary trajectory* is a trajectory  $q : [t_i, t_f] \rightarrow \mathcal{M}$  such that  $q(t_f)$  belongs to the boundary of the set of all reachable points from  $q(t_i)$

### PMP for boundary trajectories

if  $q : [t_i, t_f] \rightarrow \mathcal{M}$  is a boundary trajectory, then it has a nontrivial adjoint vector  $\psi(t)$  verifying (\*) with  $\psi_0 = 0$

### type B trajectories

type B trajectories correspond to the singularity of the control component  $u_1$  and their characterization requires geometric reasoning plus the application of PMP for boundary trajectories

**lemma** the search for optimal trajectories of type B can be restricted to the sufficient family of path types

$$l_a^+ l_b^- l_e^+ \text{ or } r_a^+ r_b^- r_e^+ \quad \text{with } 0 \leq a, b, e \leq \pi$$

## in conclusion:

sufficient family of optimal trajectories for the unicycle (PMP + geometric reasoning)

(I)	$l_a^+ l_b^- l_e^+$ or $r_a^+ r_b^- r_e^+$	$0 \leq a \leq \pi, 0 \leq b \leq \pi, 0 \leq e \leq \pi$
(II)(III)	$C_a   C_b C_e$ or $C_a C_b   C_e$	$0 \leq a \leq b, 0 \leq e \leq b, 0 \leq b \leq \pi/2$
(IV)	$C_a C_b   C_b C_e$	$0 \leq a \leq b, 0 \leq e \leq b, 0 < b \leq \pi/2$
(V)	$C_a   C_b C_b   C_e$	$0 \leq a \leq b, 0 \leq e \leq b, 0 < b \leq \pi/2$
(VI)	$C_a   C_{\pi/2} S_d C_{\pi/2}   C_b$	$0 \leq a \leq \pi/2, 0 \leq b < \pi/2, 0 \leq d$
(VII)(VIII)	$C_a   C_{\pi/2} S_d C_b$ or $C_b S_l C_{\pi/2}   C_a$	$0 \leq a \leq \pi, 0 \leq b \leq \pi/2, 0 \leq d$
(IX)	$C_a S_d C_b$	$0 \leq a \leq \pi/2, 0 \leq b \leq \pi/2, 0 \leq d$

- since  $u_1 = \pm 1$  for all the path types contained in this family, they are admissible for the Reeds and Shepp's car; this implies that the family is also sufficient for the Reeds and Shepp's time-optimal control problem
- time-optimal trajectories for the Reeds and Shepp's car are paths of minimal length (recall that for Reeds and Shepp's car  $u_1 = \pm 1$ )

## OPEN PROBLEMS

the techniques so far presented are fairly standard now, and the associated theoretical problems can be considered as solved

but: from an application viewpoint, many important issues deserve further research:

- **path planning in the presence of obstacles**: classical motion planning methods do not apply to WMRs because they ignore nonholonomic constraints
- **inclusion of dynamics**: for massive vehicles and/or at high speeds, consideration of robot dynamics is necessary for realistic control design
- **robust control design**: cope with disturbances and model perturbations (e.g., slipping)
- **use of exteroceptive feedback**: most control schemes require the measure of the WMR state; however, proprioceptive sensors, such as encoders, become unreliable in the long run  $\Rightarrow$  close the feedback loop with exteroceptive sensors providing absolute information about the robot localization in its workspace (e.g., vision)
- **WMRs not transformable in chained form**: such as a unicycle towing two or more trailers hitched at some distance from the midpoint of the previous wheel axle