

PATH AND TRAJECTORY PLANNING

viernes, 11 de octubre de 2019

11:11 a. m.

- Definitions:
 - Path in \mathcal{C}
 - $q = q(s); s \in [s_i, s_f]$
 - s : path parameter (e.g. arc length)
 - Trajectory in \mathcal{C}
 - $q = q(t); t \in [t_i, t_f]$
 - t : time
 - A trajectory can also be defined as a path and a timing law
 - $q = q(s)$
 - $s = s(t)$
 - $s(t_i) = s_i$
 - $s(t_f) = s_f$
 - This allows the problem to be divided into the **geometry** (path) of the motion and the **speed** (timing law) of the motion.
 - Path Planning**: given q_i, q_f in \mathcal{C} , find a **feasible** path (a path that the robot can follow) that goes from the initial configuration to the final one.
 - $q(s_i) = q_i$
 - $q(s_f) = q_f$
 - Trajectory Planning**: given q_i, q_f in \mathcal{C} , find a **feasible** trajectory that goes from the initial configuration to the final one over a certain time interval.
 - $q(t_i) = q_i$
 - $q(t_f) = q_f$
 - Both path and trajectory planning occur in the absence of obstacles. In the presence of obstacles the problem is called **motion planning**.
 - Direct approach**: plan directly a trajectory
 - Decoupled approach**: plan first a path and then a timing law. Convenient because the speed limits on the actuators do not limit the path, only the timing law.
 - Feasible**: WMRs are constrained by $A^T(q)\dot{q} = 0$
 - Trajectories: $q(t)$ such that:
 - $\dot{q}(t) \in N(A^T(q)) \forall t$
 - Paths:
 - $\dot{q} = \frac{dq}{dt} = \frac{\partial q}{\partial s} \frac{ds}{dt}$
 - $\frac{\partial q}{\partial s} = q'$: geometric tangent. If the s parameter is the arc length, then q' is a unit vector.
 - $\frac{ds}{dt}$: speed that modulates \dot{q} along the direction of q'
 - $A^T(q)\dot{q} = 0 \rightarrow A^T(q)q'\dot{s} = 0$
 - Since $\dot{s} \neq 0$, then:
 - $A^T(q)q' = 0$
 - Then the geometric tangents q' are constrained exactly like the \dot{q} . In general, paths that have a component along the zero motion line of a robot are not feasible.
 - Analogously to what we did with the kinematic constraints, we can represent the constraint on the geometric tangents as a linear combination of the basis of the nullspace of $A^T(q)$
 - Geometric version of the kinematic model**:

$$\square q' = \sum_{i=1}^m g_i(q) \tilde{u}_i$$

- Relation between u_i and \tilde{u}_i . Multiply by \dot{s}

$$\square q' \dot{s} = \sum_{i=1}^m g_i(q) \tilde{u}_i \dot{s}$$

$$\square u_i = \tilde{u}_i \dot{s}$$

- ◆ u_i : velocity inputs
- ◆ \tilde{u}_i : geometric inputs
- ◆ \dot{s} : speed

- **Trajectory planning:**

- Decoupled approach:
 - ◆ Choose the geometric inputs in the model
 - ◆ Integrate the model to get a path
 - ◆ Choose a timing law
- Direct approach:
 - ◆ Choose the velocity inputs in the model
 - ◆ Integrate to get a trajectory directly

- **Differential Flatness:**

- Consider a nonlinear, driftless system

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

- The system is differentially flat (DF) if there exists a set of outputs $w = h(q)$, called **flat outputs**, such that q and u can be written as algebraic functions of $(w, \dot{w}, \ddot{w}, \dots, w^{(r)})$:

$$\square q = \alpha(w, \dot{w}, \ddot{w}, \dots, w^{(r)}), u = \beta(w, \dot{w}, \ddot{w}, \dots, w^{(r)})$$

- In DF systems, knowing the evolution of the FOs allows to reconstruct the evolution of the whole state q and the history of control inputs u . **Also applies to geometric models.** So, knowing the evolution of the FOs over s allows to reconstruct the evolution of q over s (the path) and the history of u over s (the geometric inputs).
- Example: the unicycle

- Kinematic model:

$$\square \begin{pmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{pmatrix} = \begin{pmatrix} \cos \theta \\ \sin \theta \\ 0 \end{pmatrix} v + \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \omega$$

- Geometric model

$$\square \begin{pmatrix} x' \\ y' \\ \theta' \end{pmatrix} = \begin{pmatrix} \cos \theta \\ \sin \theta \\ 0 \end{pmatrix} \tilde{v} + \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \tilde{\omega}$$

- Flat outputs are:

$$\square w = \begin{pmatrix} x \\ y \end{pmatrix}$$

- x, y are known directly.

- Can we reconstruct θ ?

$$\square \theta = \text{ATAN2}(\dot{y}, \dot{x}) + k\pi; k = 0, 1$$

- The two choices for k account for the fact that the same cartesian path may be followed moving forward ($k = 0$) or backward ($k = 1$). If the initial orientation is assigned only one k is correct.

- Now that we have reconstructed all the states, we need to reconstruct the inputs.

$$\square v = \pm \sqrt{\dot{x}^2 + \dot{y}^2}$$

$$\square \omega = \dot{\theta} = \frac{1}{1 + \frac{\dot{y}^2}{\dot{x}^2}} \frac{\dot{y}\dot{x} - \dot{y}\ddot{x}}{\dot{x}^2} = \frac{(\ddot{y}\dot{x} - \dot{y}\ddot{x})}{\dot{x}^2 + \dot{y}^2}$$

- If $\dot{x} = \dot{y} = 0$ then we cannot reconstruct the ω or θ .
- Interpretation: We have a movie of the unicycle motion from the ceiling, so we see its position and its motion but θ is hidden. By knowing the motion of the center we can reconstruct the whole state and inputs. If the robot is not moving we cannot reconstruct θ and ω
- Exercise: Bicycle.

- Kinematic model

$$\square \begin{pmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \\ \dot{\phi} \end{pmatrix} = \begin{pmatrix} \cos \theta \\ \sin \theta \\ \frac{\tan \phi}{l} \\ 0 \end{pmatrix} v + \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \omega$$

- Geometric model

$$\square \begin{pmatrix} x' \\ y' \\ \theta' \\ \phi' \end{pmatrix} = \begin{pmatrix} \cos \theta \\ \sin \theta \\ \frac{\tan \phi}{l} \\ 0 \end{pmatrix} \tilde{v} + \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \tilde{\omega}$$

- Flat outputs

$$\square w = \begin{pmatrix} x \\ y \end{pmatrix}$$

- Reconstruct θ and ϕ

$$\square \theta = \text{ATAN2}(\dot{y}, \dot{x}) + k\pi; k = 0, 1$$

$$\square \phi = \text{ATAN2}(l\omega, v)$$

- But

$$\square v = \pm \sqrt{\dot{x}^2 + \dot{y}^2}$$

$$\square \omega = \dot{\theta} = \frac{(\ddot{y}\dot{x} - \dot{y}\ddot{x})}{\dot{x}^2 + \dot{y}^2}$$

- It is important to notice that the value of ϕ changes with the choice of the velocity input v

- Example: the $(2, n)$ chained form

- Flat outputs:

$$\square w = \begin{pmatrix} z_1 \\ z_n \end{pmatrix}$$

- Reconstruct state:

$$\square z_{k-1} = \dot{z}_k / \dot{z}_1$$

$$\square \text{For } k \in [2, n]$$

- Reconstruct inputs:

$$\square v_1 = \dot{z}_1$$

$$\square v_2 = \dot{z}_2$$

$$\square \dot{z}_{k-1} = \frac{\ddot{z}_k \dot{z}_1 - \dot{z}_k \ddot{z}_1}{\dot{z}_1^2}$$

$$\square \text{For } k \in [2, n]$$

- If a system is DF then it can be expressed in chained form and viceversa.

• Path Planning:

- For path planning we consider the geometric version of the kinematic model:

$$\square q' = \sum_{i=1}^m g_i(q) \tilde{u}_i$$

- $q' = \frac{\partial q}{\partial s}$: geometric tangent
- $\tilde{u}_i(s) = \frac{u_i(t)}{s}$: geometric inputs

- **Scheme 1:** With **Differential Flatness** (and parametrized paths): **If the system is flat then we can describe the whole evolution of the state and the inputs from the FOs.**

a. Compute boundary conditions

- $w_i = h(q_i)$
- $w_f = h(q_f)$

b. Generate a path for the FOs $w \in [w_i, w_f]$: using a geometric path in s .

c. Reconstruct the path for the whole state $q(s)$ and for the geometric inputs $\tilde{u}_i(s)$ using the reconstruction formulas of flatness.

○ Example: Unicycle

- $q_i = \begin{pmatrix} x_i \\ y_i \\ \theta_i \end{pmatrix} \rightarrow q_f = \begin{pmatrix} x_f \\ y_f \\ \theta_f \end{pmatrix}$

i. Boundary conditions $s \in [0,1]$:

- $w_i = \begin{pmatrix} x_i \\ y_i \end{pmatrix} w_f = \begin{pmatrix} x_f \\ y_f \end{pmatrix}$
- $x'(0) = \tilde{v}(0) \cos \theta_i$
- $y'(0) = \tilde{v}(0) \sin \theta_i$
- $x'(1) = \tilde{v}(1) \cos \theta_f$
- $y'(1) = \tilde{v}(1) \sin \theta_f$
- The initial and final velocities are **free** parameters.

ii. Generate a path for x and y

- Cubic polynomials (we can choose any interpolating basis that satisfies the constraints)
 - ◆ $x(s) = a_x s^3 + b_x s^2 + c_x s + d_x$
 - ◆ $y(s) = a_y s^3 + b_y s^2 + c_y s + d_y$
- We get:
 - ◆ $x(s) = s^3 x_f - (s-1)^3 x_i + \alpha_x s^2 (s-1) + \beta_x s (s-1)$
 - ◆ $y(s) = s^3 y_f - (s-1)^3 y_i + \alpha_y s^2 (s-1) + \beta_y s (s-1)$
 - ◆ $\begin{pmatrix} \alpha_x \\ \alpha_y \end{pmatrix} = \begin{pmatrix} K \cos \theta_f - 3x_f \\ K \sin \theta_f - 3y_f \end{pmatrix}$
 - ◆ $\begin{pmatrix} \beta_x \\ \beta_y \end{pmatrix} = \begin{pmatrix} K \cos \theta_i - 3x_i \\ K \sin \theta_i - 3y_i \end{pmatrix}$
 - ◆ $v(0) = v(1) = K$: this is a free choice, they do not have to be equal.

iii. Reconstruct the path:

- $\theta(s) = \text{ATAN2}(y'(s), x'(s))$
- $\tilde{v}(s) = \pm \sqrt{x'^2 + y'^2}$
- $\tilde{\omega}(s) = \frac{(y''x' - y'x'')}{x'^2 + y'^2}$

• **Scheme 2: With chained form (and parametrized inputs): if the robot can be transformed in chained form.**

- This technique can be easily generalized to any kinematic model.
- This approach consists of parametrizing the inputs instead of the path.
- Geometric version of the chained form:

- $\dot{z}_1' = \tilde{v}_1$
- $\dot{z}_2' = \tilde{v}_2$
- $\dot{z}_3' = z_2 \tilde{v}_1$
- ...
- $\dot{z}_n' = z_{n-1} \tilde{v}_1$

- First we have to convert to the chained form the initial and desired configurations, So we need to transform the coordinates from q_i, q_f to z_i, z_f . With the hypothesis:

- $z_{1,f} \neq z_{1,i}$
- $\Delta = z_{1,f} - z_{1,i}$

a. $\tilde{v}_1 = \text{sign}(\Delta)$

- $\text{sign}(\Delta) = 1; \Delta > 0$

- $sign(\Delta) = 0; \Delta = 0$
- $sign(\Delta) = -1; \Delta < 0$
- b. $\tilde{v}_2 = c_0 + c_1 s + \dots + c_{n-2} s^{n-2}$
 - $n - 1$ free parameters to impose the $n - 1$ conditions
 - With $s \in [s_i, s_f] = [0, |\Delta|]$
 - $z_2(|\Delta|) = z_{2,f}$
 - $z_n(|\Delta|) = z_{n,f}$
 - Exploiting the fact that the system is linear, \tilde{v}_1 is constant.
 - This results in a linear system of equations:

$$\square D(z_f, \Delta) \begin{pmatrix} c_0 \\ c_1 \\ \dots \\ c_{n-2} \end{pmatrix} = d(z_f, \Delta)$$

- ◆ D : always non singular if $\Delta \neq 0$
- ◆ c_i : unknowns

- c. Knowing \tilde{v}_1, \tilde{v}_2 , integrate the system to obtain z_i . Then use inverse transformation to get original coordinates q and inputs u

- Remarks:

- Both schemes are guaranteed to produce **feasible** paths.
 - Scheme 1: uses reconstruction formulas that are based on the model (geometric)
 - Scheme 2: paths are obtained by direct integration the model (geometric)
 - All generated paths will automatically satisfy the constraints (encoded in the model)
- All systems that are flat can be represented in chained form and viceversa.
 - Scheme 1: requires differential flatness
 - Scheme 2: requires chained form transformability

- **Trajectory Planning:**

- **Decoupled approach:** we have a path $q(s), s \in [s_i, s_f]$. We want to add a timing law $s = s(t), t \in [t_i, t_f]$, with $s(t_i) = s_i, s(t_f) = s_f$.
 - a. We need to choose any $s(s)$ so that:
 - $u_i = \tilde{u}_i \dot{s}$
 - b. Check if the velocity bounds are satisfied with the chosen $s(t)$
 - $|u_i(t)| \leq u_{i,max}$
 - c. If not, redefine the time:
 - $\tau = \frac{t}{T}$
 - $\dot{s} = \frac{ds}{dt} = \frac{ds}{d\tau} \frac{d\tau}{dt} = \frac{ds}{d\tau} \frac{1}{T}$
- **Direct approach:** use scheme 1 or 2 with t in place of s using the kinematic model in place of the geometric model. Applying the velocity bounds on this approach not only changes the timing law while keeping the path, **it will change the whole trajectory**, i.e. changing both the path and the timing law.

Chained Form

A "canonical" form for kinematic models of WMR

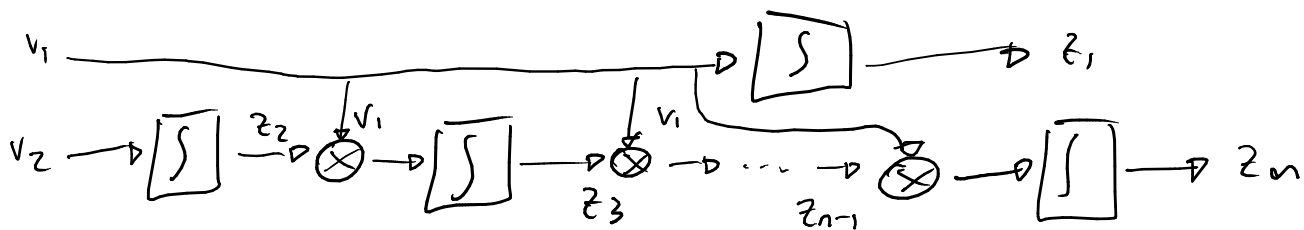
$$\dot{z} = \begin{pmatrix} \dot{z}_1 \\ \vdots \\ \dot{z}_n \end{pmatrix} = \begin{pmatrix} v_1 \\ v_2 \\ z_2 v_1 \\ \vdots \\ z_{n-1} v_1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ z_2 \\ \vdots \\ z_{n-1} \end{pmatrix} \overset{\gamma_1}{v_1} + \begin{pmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \overset{\gamma_2}{v_2}$$

A particular driftless 2-input dynamical system

$z \in \mathbb{R}^n$ STATE

$v \in \mathbb{R}^2$ inputs

Block scheme



Controllable

$$[\gamma_1, \gamma_2], [\gamma_1, [\gamma_1, \gamma_2]], \dots \rightarrow \dim \Delta_A = n \text{ (verify!)}$$

There exist NCS conditions for transforming

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

$$\text{into } \dot{z} = \gamma_1(z)v_1 + \gamma_2(z)v_2$$

all systems like (*) with $n \leq 4$ satisfy these conditions

(unicycle bicycle can be put in chained form)
 * tricycle may not!

How to transform?

via

(1) coord transformation $z = \alpha(q)$

(2) input transformation $v = \beta(q)u$

↳ a feedback transformation (it needs the state of the robot)

unicycle

$$\begin{cases} \dot{x} = v \cos \theta \\ \dot{y} = v \sin \theta \\ \dot{\theta} = \omega \end{cases} \quad (1) \quad v.s. \quad \begin{cases} \dot{z}_1 = v_1 \\ \dot{z}_2 = v_2 \\ \dot{z}_3 = z_2 v_1 \end{cases} \quad \text{chained form} \quad (2)$$

$$q = \begin{pmatrix} x \\ y \\ \theta \end{pmatrix} \quad u = \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \begin{pmatrix} v \\ \omega \end{pmatrix} \quad \text{drift vel.}$$

(1) becomes (2) letting

$$\begin{cases} z_1 = \theta \\ z_2 = x \cos \theta + y \sin \theta \\ z_3 = x \sin \theta - y \cos \theta \end{cases}$$

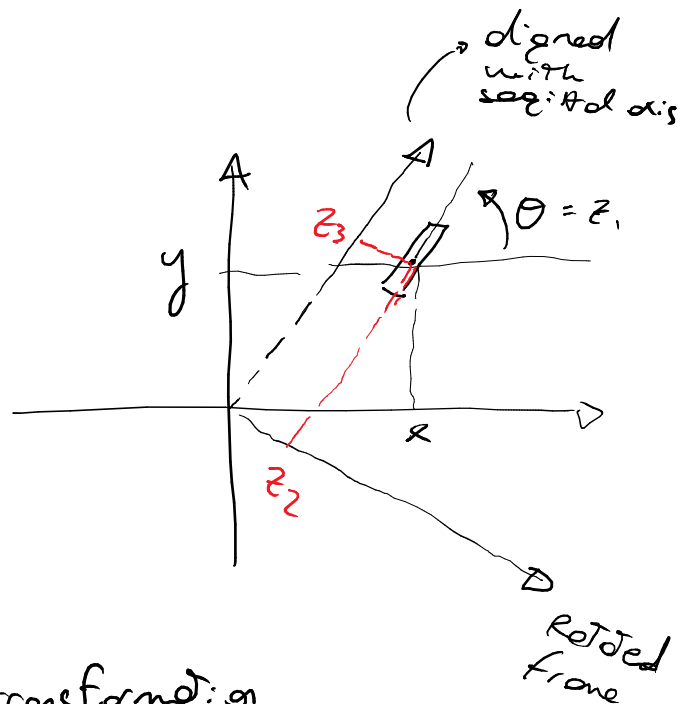
↓
(2,3) chained form
↓ states

inputs

$$\boxed{\dot{z}_1 = \dot{\theta} = \omega = v_1} \quad \text{1st input transformation}$$

$$\begin{aligned} \dot{z}_2 &= \dot{x} \cos \theta - x \sin \theta \dot{\theta} + \dot{y} \sin \theta - y \cos \theta \dot{\theta} \\ &= v \cos^2 \theta + v \sin^2 \theta - (x \sin \theta + y \cos \theta) \dot{\theta} \\ &= v - z_2 \omega = v_2 \end{aligned} \quad \text{2nd input transformation}$$

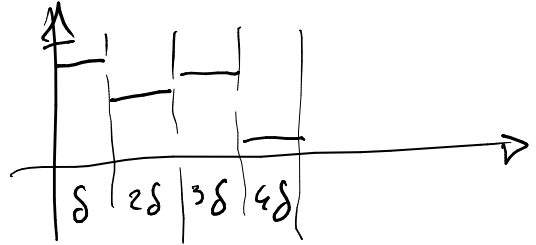
$$\dot{z}_3 \text{ to check using coord. \& input transf.} = z_2 v_1$$



Use of chained form

good for planning & control

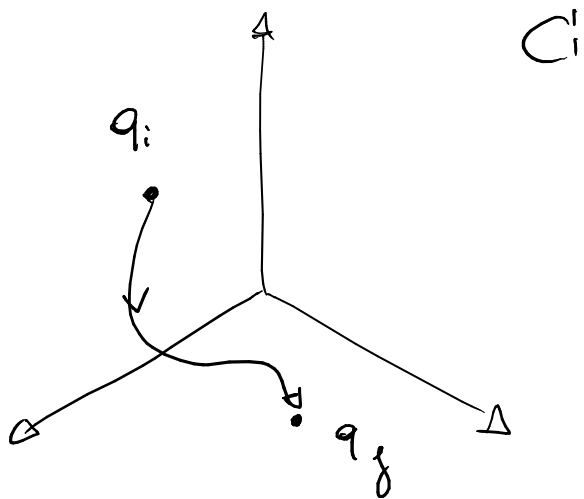
- It applies to a large number of WTR
- It can be easily integrated under appropriate inputs (e.g. piecewise constant)



Path & Trajectory planning

→ for WER

- Path, trajectory
- Timing law
- Differential flatness
- Path planning
- Trajectory planning



path $q(s), s \in [s_i, s_f]$
 $q(s_i) = q_i, q(s_f) = q_f$ I can go back & forth with s on the path
path parameter (typically the arc length)

trajectory $q(t), t \in [t_i, t_f]$
 $q(t_i) = q_i, q(t_f) = q_f$ → particular path with time as parameter, I can't go back

Path: Geometry & motion

Trajectory: Geometry + speed motion

$q = q(t), t \in [t_i, t_f]$ can be separated in

① $q = q(s), s \in [s_i, s_f]$
 $q(s_i) = q_i, q(s_f) = q_f$

path

+

② $s = s(t), t \in [t_i, t_f]$
 $s(t_i) = s_i, s(t_f) = s_f$

timing law

Path Planning

given $q_i, q_f \in C$ find a path $q = q(s), s \in [s_i, s_f]$
such that $q(s_i) = q_i$
 $q(s_f) = q_f \quad \rightarrow \quad \text{A geometric Problem}$

Trajectory planning

Given $q_i, q_f \in C$, find a trajectory $q(t), t \in [t_i, t_f]$
such that $q(t_i) = q_i$
 $q(t_f) = q_f \quad \rightarrow \quad \text{A kinematic Problem}$

Usually path & trajectory are defined with the absence of obstacles (treated in motion planning)

2 approaches

- ① all-in-one : directly plan a trajectory $q(t)$
- ② decoupled (two phases) : first plan a path then a timing law (better than ① because we separate the plan part from the timing law)

Non holonomic constraints

→ local mobility is restricted:

$$\dot{q} \in N(A^T(q))$$

↳ not all generalized velocities are admissible

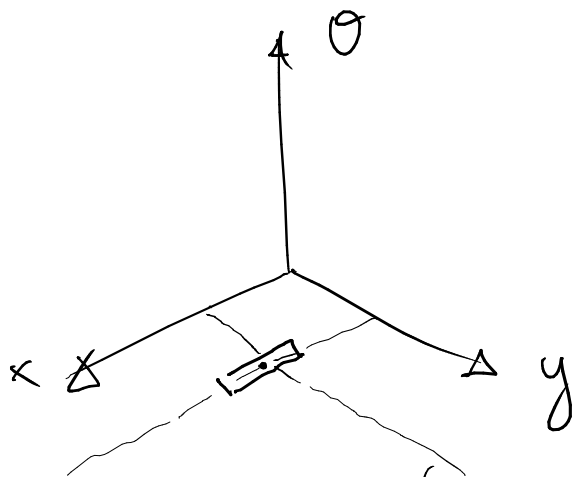
therefore

not all trajectories in Q are admissible!

• What about paths?

Also paths are restricted cause it's a matter of geometry

example: unicycle



↳ not a possible trajectory (ZML)



the path along the ZML is actually not admissible

Analytically

\dot{q} what are these generalized velocities?

$$\dot{q} = \frac{\partial q}{\partial t}$$

$$\begin{pmatrix} \dot{q}_1 \\ \vdots \\ \dot{q}_n \end{pmatrix}$$

$q = q(t)$ a trajectory seen as

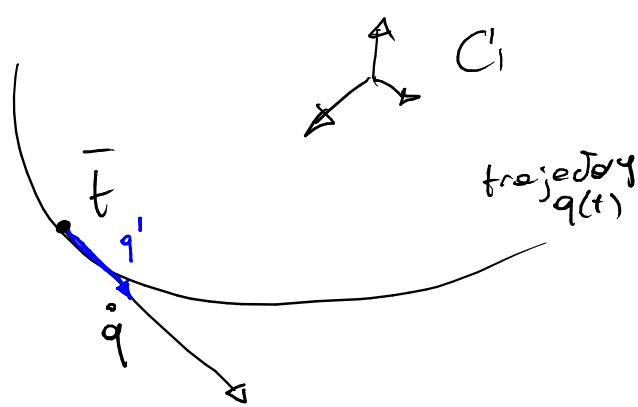
$$q = q(s(t))$$

therefore $\dot{q} = \dot{q}' \cdot \dot{s}$

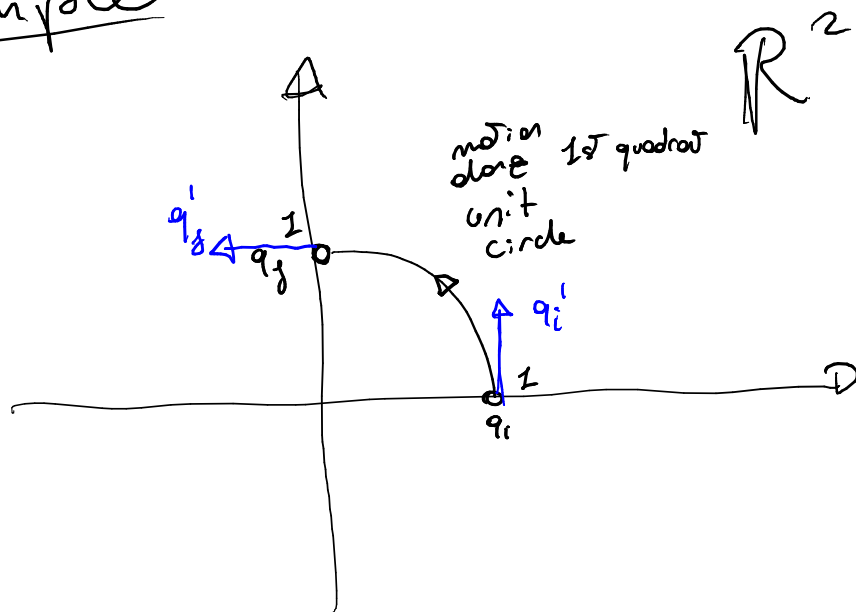
$$\frac{\partial q}{\partial t} = \frac{\partial q}{\partial s} \frac{\partial s}{\partial t} = q' \cdot \dot{s}$$

$$\dot{q} \cdot \dot{s} \leadsto \text{a scalar}$$

$$\begin{pmatrix} \frac{\partial q_1}{\partial s} \\ \vdots \\ \frac{\partial q_n}{\partial s} \end{pmatrix} \text{ the tangent vector is the path at } s$$



example



$$q(s) = \begin{pmatrix} \cos s \\ \sin s \end{pmatrix} \quad s \in [0, \pi/2] \quad s: \text{arc length}$$

$$q' = \begin{pmatrix} -\sin s \\ \cos s \end{pmatrix} \text{ the tangent vector}$$

$$\|q'\| = 1 \quad \text{because } s \text{ is the arc length!}$$

exercise : find a different parametrization (e.g. $x=s$)
compute q'
verify it's a tangent vector, $\|q'\| \neq 1$

↳ because we are not using arc length

$$\dot{q} = q' \cdot \dot{s} \quad \text{then}$$

$A^T(q) \dot{q} = 0$ position constraints can be written as

$$A^T(q) q' \dot{s} = 0$$

\hookrightarrow must be true
for any \dot{s}



$$A^T(q) q' = 0$$

Kinematic constraints actually
are constraints on tangent vectors
to the path

let's exploit this part. \rightarrow

$\{e_1(q), \dots, e_n(q)\}$ a basis of the null space $N(A^T(q))$
 same as before

$$(*) \quad \dot{q} = \sum_{j=1}^n \underbrace{e_j(q)}_{\substack{\text{same vgs} \\ \text{as before}}} \tilde{u}_j$$

different coefficients \rightarrow different inputs

time derivatives replaced by derivatives wrt. s

\rightarrow Geometric version of the kinematic model

Relationship u_j / \tilde{u}_j

\downarrow
 velocity inputs \rightarrow geometric inputs

multiply (*) by \dot{s}

$$\begin{aligned} \dot{q} \cdot \dot{s} &= \sum_{j=1}^n e_j(q) \tilde{u}_j \cdot \dot{s} \\ &= \dot{q} = \sum_{j=1}^n e_j(q) \underbrace{\tilde{u}_j}_{u_j} \cdot \dot{s} \end{aligned}$$

$$\Rightarrow \boxed{\begin{matrix} \text{velocity inputs} & \text{geometric inputs} \\ u_j & = \tilde{u}_j \cdot \dot{s} \end{matrix}}$$

interpretation of (*)

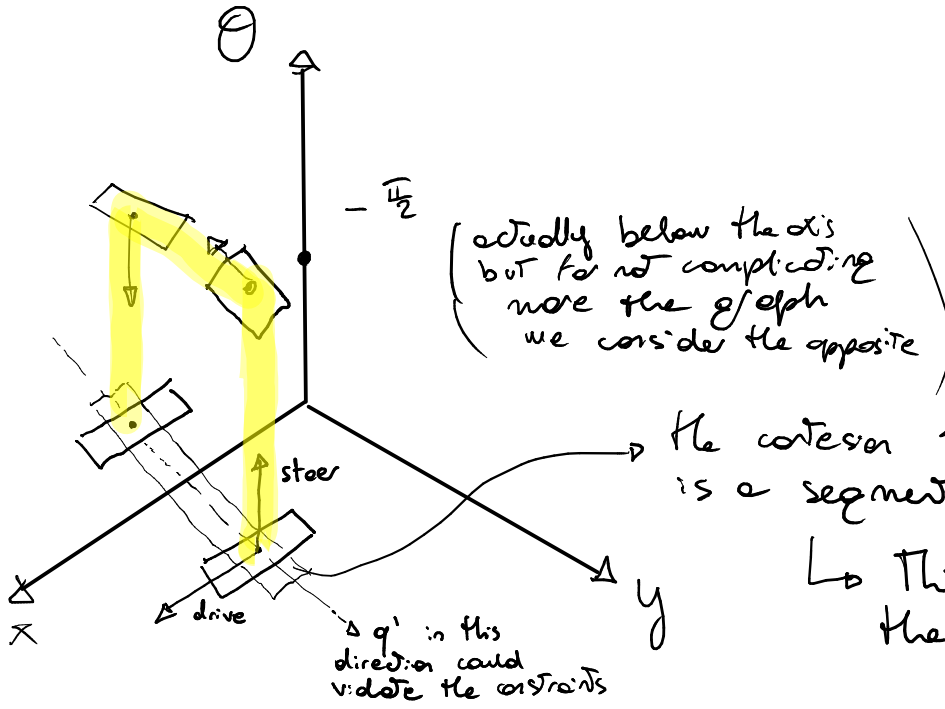
\rightarrow kinematic model:
 Good for trajectory planning

Choose any \tilde{u}_j ($j=1 \dots n$), the model will predict the resulting path

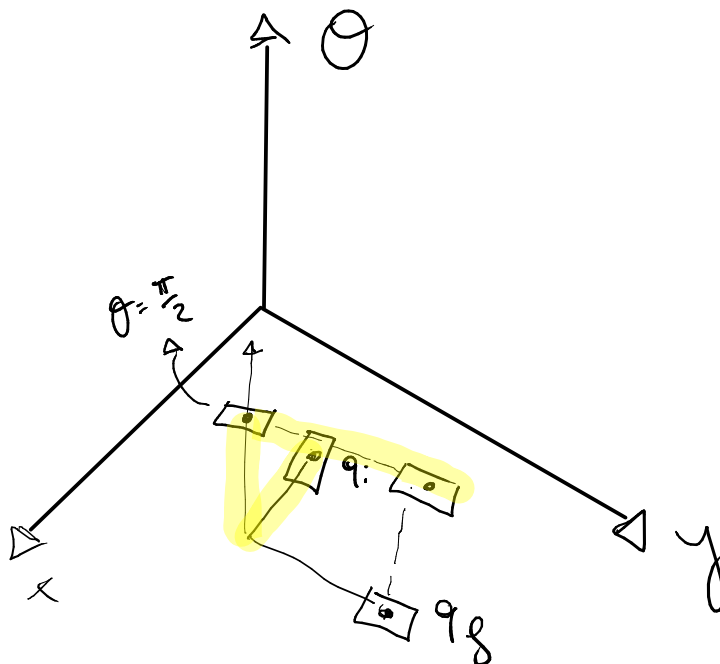
\rightarrow good for path planning

Unicycle

$$\dot{q} = \begin{pmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{pmatrix} = \underbrace{\begin{pmatrix} \cos \theta \\ \sin \theta \\ 0 \end{pmatrix}}_{\text{drive}} \tilde{v} + \underbrace{\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}}_{\text{steer}} \tilde{\omega}$$



↳ This, however, is NOT the C-space motion!



Differentially flatness

A system is differentially flat if states and inputs can be RECONSTRUCTED from the evolution of some outputs (FLAT outputs)

Definition

$$\dot{q} = \sum_{j=1}^n e_j(q) u_j = G(q)u \quad (\text{or } \dot{q} = \sum_{j=1}^n e_j(q) \tilde{u}_j = G(q)\tilde{u})$$

is differentially flat if there exist a set of outputs

$W = h(q)$ such that

W = flat outputs

$$\begin{cases} q = \eta(W, \dot{W}, \dots, W^{(r)}) \\ u = \mu(W, \dot{W}, \dots, W^{(r)}) \end{cases}$$

\Rightarrow these two allow an ALGEBRAIC RECONSTRUCTION OF STATE / INPUTS



We don't need to integrate then we don't need initial conditions.

$$\eta(W, \dot{W}, \dots, W^{(r)})$$

$$\mu(W, \dot{W}, \dots, W^{(r)})$$

\leadsto if we are considering a geometric model

Unicycle

$$\begin{cases} \dot{x} = v \cos \theta \\ \dot{y} = v \sin \theta \\ \dot{\theta} = \omega \end{cases}$$

is differentially flat

flat outputs $W = \begin{pmatrix} x \\ y \end{pmatrix}$

• feedback: state reconstruction ($W \rightarrow q$)

$$\begin{cases} \dot{x} = W_1 = x \\ \dot{y} = W_2 = y \\ \dot{\theta} = \arctan \frac{\dot{y}}{\dot{x}} \end{cases}$$

• input reconstruction ($W \rightarrow v, \omega$)

$$\begin{cases} v = \pm \sqrt{\dot{x}^2 + \dot{y}^2} \begin{matrix} \swarrow + \text{(forward motion)} \\ \searrow - \text{(backward motion)} \end{matrix} \\ \omega = \frac{1}{1 + \left(\frac{\dot{y}}{\dot{x}}\right)^2} \frac{\ddot{y}\dot{x} - \ddot{x}\dot{y}}{\dot{x}^2} = \frac{\ddot{y}\dot{x} - \ddot{x}\dot{y}}{\dot{x}^2 + \dot{y}^2} \end{cases}$$

$$= \text{ATAN2}(\dot{y}, \dot{x}) + K\pi$$

$K=0$ if $v > 0$
 $K=1$ if $v < 0$

For the cartesian trajectory

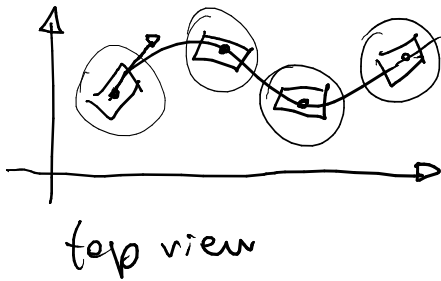
$x(t), y(t)$

↳ state trajectory $q(t)$
input history $u(t)$

The possibility of reconstruction is lost when $\dot{x}=\dot{y}=0$
(motion degenerates to a point)

interpretation

⊙ snapshot



need a move for reconstruction

↳ cannot tell θ (state reconstruction is lost)

cannot tell w (input reconstruction is lost)

Path & Trajectory Planning

differential flatness

$$w = h(q) \quad \text{flat outputs}$$

\downarrow
function of the state

$$\begin{cases} q = q(w, \dot{w}, \dots, w^{(r)}) \\ v = v(w, \dot{w}, \dots, w^{(r)}) \end{cases} \quad \text{ALGEBRAIC reconstruction}$$

Unicycle : flat outputs are x, y

Bicycle : flat outputs are x, y

Tricycle with Trailer : //

Chained form also differentially flat

$$\dot{z}_1 = v_1$$

$$\dot{z}_2 = v_2$$

$$\dot{z}_3 = z_2 v_1$$

(2,3) case

flat outputs, $\underline{z}_1, \underline{z}_3$

$$z_2 = \frac{\dot{z}_3}{v_1} = \frac{\ddot{z}_3}{\dot{z}_1}$$

state reconstruction:
algebraic
function of the
1st order derivative

$$v_1 = \dot{z}_1$$

$$v_2 = \dot{z}_2 = \frac{\ddot{z}_3 \dot{z}_1 - \dot{z}_3 \ddot{z}_1}{\dot{z}_1^2}$$

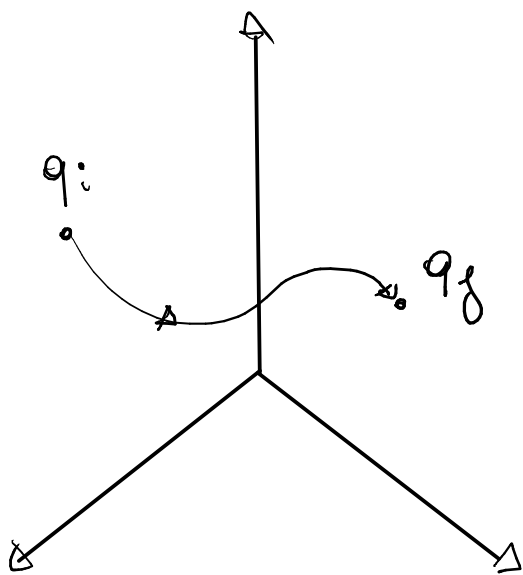
(2,n) chained form $\rightarrow z_1, z_n$ are FLAT outputs

Use of flatness for planning

flow outputs can evolve arbitrarily

\Rightarrow Then the reconstruction formulas will provide the associated state trajectory, which will be automatically feasible

Generic algorithm for path planning / trajectory plan. using flatness



(hyp)

what is flow

$$w = h(q)$$

↑
flow
outputs

1. Compute $w_i = h(q_i)$, $w_f = h(q_f)$
2. Generate a path (in s) or a trajectory (in t) from w_i to w_f with the appropriate boundary conditions (interpolation problem)
3. Use the reconstruction formulas to compute the state and input

↓
path / trajectory

example : unicycle

$$q_i = \begin{pmatrix} x_i \\ y_i \\ \theta_i \end{pmatrix} \rightarrow q_f = \begin{pmatrix} x_f \\ y_f \\ \theta_f \end{pmatrix}$$

$$w_i = \begin{pmatrix} x_i \\ y_i \end{pmatrix} \quad w_f = \begin{pmatrix} x_f \\ y_f \end{pmatrix}$$

$$\begin{cases} x' = \tilde{v} \cos \theta \\ y' = \tilde{v} \sin \theta \\ \theta' = \tilde{\omega} \end{cases} \Rightarrow \text{Geometric model} \rightarrow \begin{matrix} \text{path} \\ \text{planning} \\ (\text{not} \\ \text{trajectory}) \end{matrix}$$

boundary conditions

$$\begin{aligned} x'(s_i) &= \tilde{v}(s_i) \cos \theta_i \\ y'(s_i) &= \tilde{v}(s_i) \sin \theta_i \\ x'(s_f) &= \tilde{v}(s_f) \cos \theta_f \\ y'(s_f) &= \tilde{v}(s_f) \sin \theta_f \end{aligned} \quad \begin{matrix} \rightarrow \kappa_i \\ \leftarrow \kappa_f \end{matrix}$$

x must go from x_i to x_f satisfying $x'(s_i) = \dots (\#)$
and the same for $y \dots$ $x'(s_f) = \dots$

\Rightarrow a cubic polynomial
for $s \in [0, 1] = [s_i, s_f]$

$$x(s) = s^3 \alpha_x - (s-1)^3 x_i + \alpha_x s^2 (s-1) + \beta_x s (s-1)$$

$$x(0) = x_i$$

$$x(1) = x_f$$

α_x, β_x so as to satisfy (#)

$$\alpha_x = \kappa \cos \theta_f - 3x_f \quad (\kappa = \kappa_i = \kappa_f)$$

$$\beta_x = \kappa \cos \theta_i + 3x_i$$

And the same for y

$$\begin{aligned} x(s), x'(s), x''(s) \\ y(s), y'(s), y''(s) \end{aligned} \xrightarrow{\text{reconst.}} \theta(s), v(s), \omega(s)$$

example : chained form

$(2, 3)$

$$\begin{cases} z_1' = \tilde{v}_1 \\ z_2' = \tilde{v}_2 \\ z_3' = z_2 \tilde{v}_1 \end{cases} \quad (\text{geometric model})$$

$$z_i = \begin{pmatrix} z_{1i} \\ z_{2i} \\ z_{3i} \end{pmatrix} \rightarrow z_f = \begin{pmatrix} z_{1f} \\ z_{2f} \\ z_{3f} \end{pmatrix}$$

$$w_i = \begin{pmatrix} z_{1i} \\ z_{3i} \end{pmatrix} \rightarrow w_f = \begin{pmatrix} z_{1f} \\ z_{3f} \end{pmatrix}$$

plot outputs : z_1, z_3

(see book for interpolation)

boundary conditions

$$z_2 = \frac{z_3'}{z_1'} \rightarrow \left(\frac{z_3'(s_f)}{z_1'(s_f)} = z_{2f} \text{ and } \frac{z_3'(s_i)}{z_1'(s_i)} = z_{2i} \right)$$

An alternative approach (not based on differential flatness) works for robots that can be transformed into desired form

idea: easily

path planning: easily integrable with the appropriate inputs

$$\begin{aligned} z_1' &= \tilde{v}_1 \\ z_2' &= \tilde{v}_2 \\ z_3' &= z_2 \tilde{v}_1 \\ &\vdots \\ z_n' &= z_{n-1} \tilde{v}_1 \end{aligned}$$

define $\Delta = z_{1,j} - z_{1,i}$, assume $\Delta \neq 0$

choose $\tilde{v}_1 = \text{sgn}(\Delta) = (\pm 1)$

\downarrow

$$\tilde{v}_2 = \underbrace{C_0}_{s_i} + \underbrace{C_1}_{s_g} s + \dots + \underbrace{C_{n-2}}_{s_g} s^{n-2} \quad \begin{array}{l} n-1 \text{ params} \\ \text{to "place" } n-1 \\ \text{variables} \\ z_2, \dots, z_{n-1} \end{array}$$

$$s \in [0, |\Delta|]$$

Integrate (can be done because $\tilde{v}_1 = \pm 1$) the desired form equations and impose that

$$\begin{cases} z_2(1|\Delta|) \stackrel{!}{=} z_{2,j} \\ \vdots \\ n-1 \text{ eqs in } n-2 \text{ unknowns} \rightarrow \text{a linear system} \\ \vdots \\ z_n(1|\Delta|) \stackrel{!}{=} z_{n,j} \end{cases}$$

$$\mathcal{D}(\Delta) \begin{pmatrix} C_0 \\ C_1 \\ \vdots \\ C_{n-2} \end{pmatrix} = b(z_i, z_j, \Delta)$$

unique iff $\Delta \neq 0$

e.g. in the (2,3) case

$$\mathcal{D}(\Delta) = \begin{pmatrix} |\Delta| & \Delta^2/2 \\ \text{sgn}(\Delta) \frac{\Delta^2}{2} & \frac{\Delta^3}{6} \end{pmatrix}$$

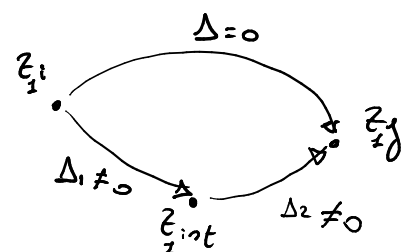
$$b(\dots) = \begin{pmatrix} z_{2,j} - z_{2,i} \\ z_{3,j} - z_{3,i} - z_{2,i} \Delta \end{pmatrix}$$

if $\Delta = 0 \rightarrow \Delta = z_{1,j} - z_{1,i}$

1. add a "via point"

2. if z_i is on circle: (as for the unicycle)

$$\text{let } z_{1,j}^{\text{new}} = z_{1,j}^{\text{old}} + 2\pi$$

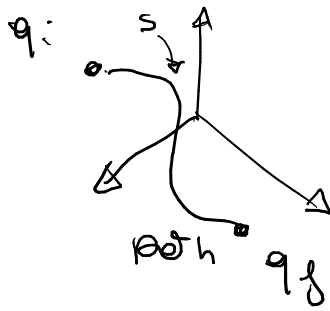


Choice of timing law

We have a path $q(s) \in [s_i, s_f]$

$$q(s_i) = q_i$$

$$q(s_f) = q_f$$



how do we choose a timing law $s = s(t)$?

A trajectory is needed to actually control the robot in time

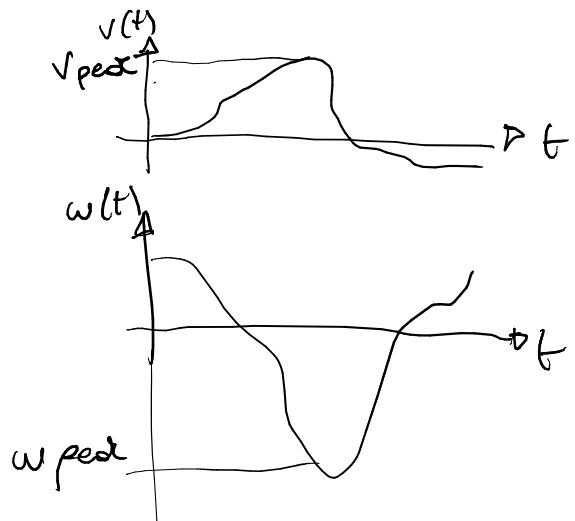
velocity bounds

unicycle $|v(t)| \leq v_{\max}$
 $|\omega(t)| \leq \omega_{\max}$

consider these bounds when choosing a timing law

let $t = \alpha s$ ($s = \frac{1}{\alpha}$)

1. $\alpha = 1$, compute $v(t)$
 $\omega(t)$



2. if $|v_{\text{peak}}| \leq v_{\max}$
 $|\omega_{\text{peak}}| \leq \omega_{\max}$

Then $\alpha = 1$ is ok $\Rightarrow s = t$

3. else, let $\eta = \max\left(\frac{|v_{\text{peak}}|}{v_{\max}}, \frac{|\omega_{\text{peak}}|}{\omega_{\max}}\right)$ and let $\alpha = \frac{1}{\eta}$
 \Rightarrow new $|v| \leq v_{\max}$ & $|\omega| \leq \omega_{\max}$ with (at least) one velocity saturating the bound at peak value

Autonomous and Mobile Robotics

Prof. Giuseppe Oriolo

Wheeled Mobile Robots 3 Path/Trajectory Planning

companion slides for the blackboard lecture

DIPARTIMENTO DI INGEGNERIA INFORMATICA
AUTOMATICA E GESTIONALE ANTONIO RUBERTI

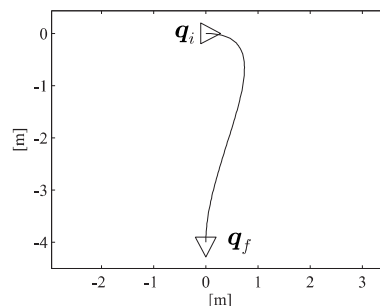
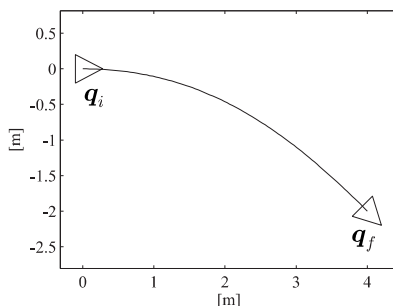


SAPIENZA
UNIVERSITÀ DI ROMA

parking a unicycle: numerical results

I. forward parking

cubic polynomials for cartesian coords x, y (flat outputs)

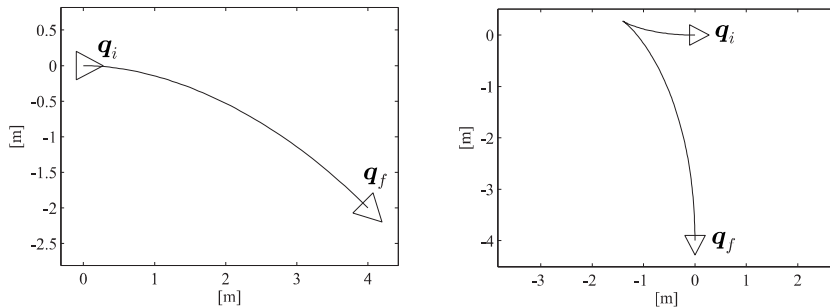


- $k=5>0$, hence forward motion
- no motion inversions

parking a unicycle: numerical results

I. forward parking

parameterized inputs on chained form

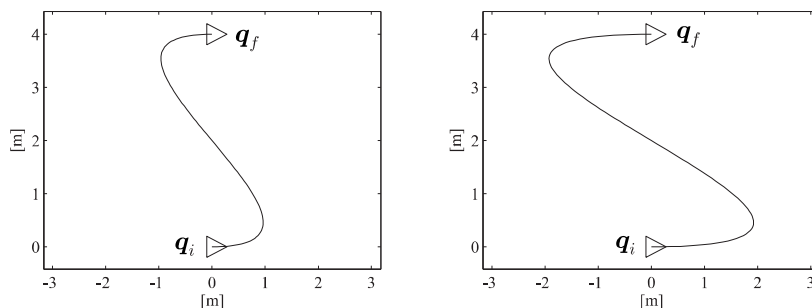


- first maneuver is similar
- a motion inversion (cusp) in the second

parking a unicycle: numerical results

2. parallel parking

cubic polynomials for cartesian coords x, y (flat outputs)

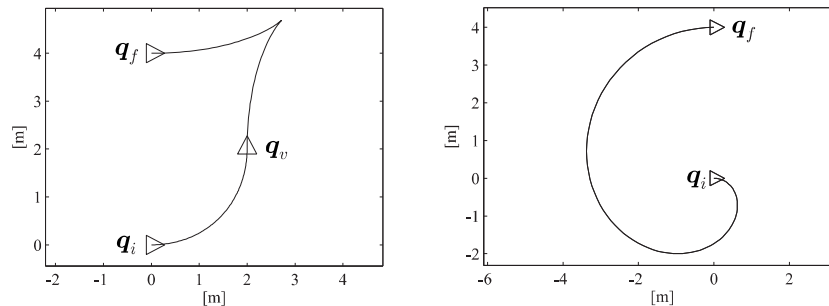


- left: $k=10$, right: $k=20$
- no motion inversions

parking a unicycle: numerical results

2. parallel parking

parameterized inputs on chained form

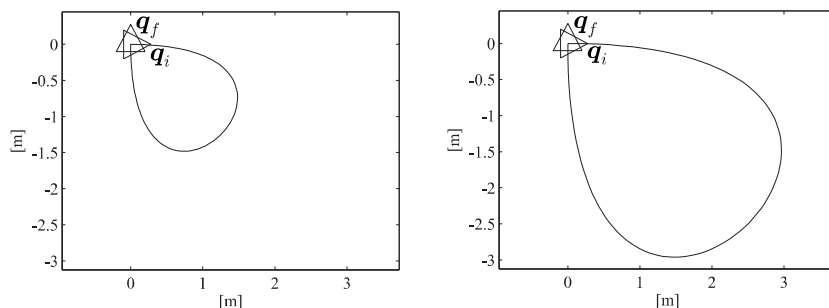


- left: with a via point
- right: requiring a full rotation

parking a unicycle: numerical results

3. pure reorientation

cubic polynomials for cartesian coords x, y (flat outputs)

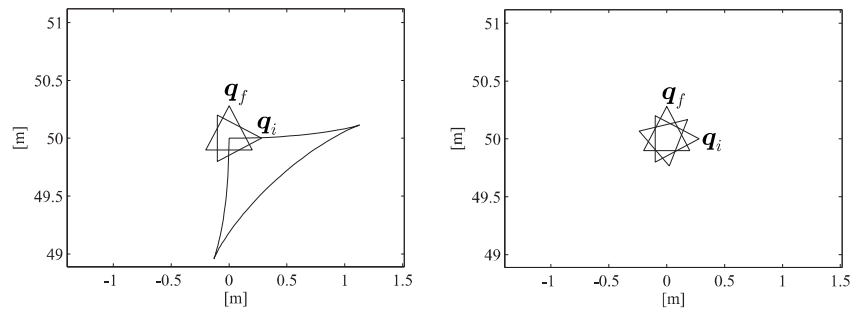


- left: $k=10$, right: $k=20$
- need to move the cartesian coordinates!

parking a unicycle: numerical results

3. pure reorientation

parameterized inputs on chained form



- left: straightforward
- right: placing the origin of z_2, z_3 at q_i