Applicazioni dell'Automatica

Introduction to mobile robotics Kinematics and modeling of WMRs

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outline of this lecture

- ground locomotion and balance
- wheels
- kinematic structures
- unicycle robot and its kinematic model
- nonholonomic constraints
- unicycle: equivalent vehicles
- car-like robot and its kinematic model
- unicycle vs. car-like model: mobility
- controllability
- odometric localization

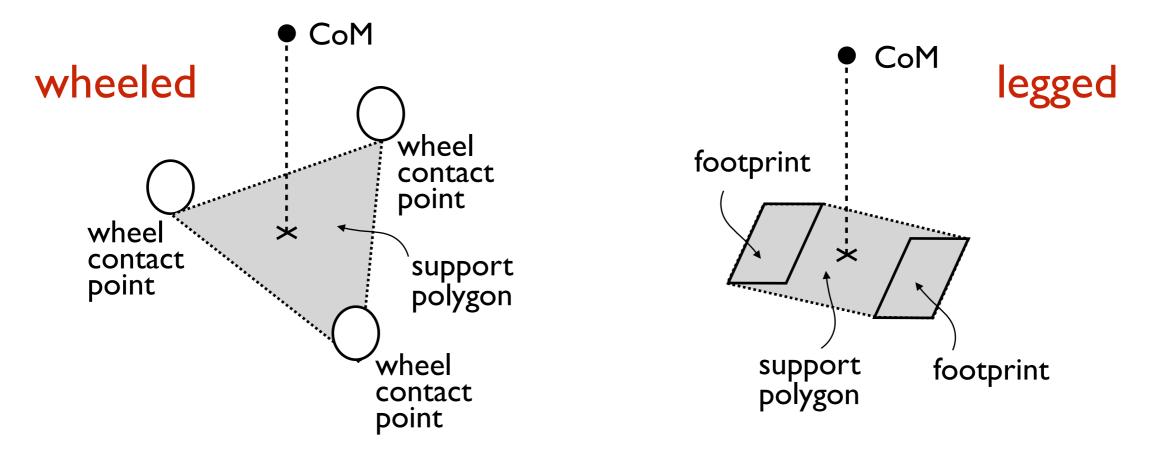
ground locomotion

- requires contact via
 - wheels: wheeled mobile robots (WMRs), typically consisting of a rigid body (base or chassis) + wheels
 - feet: legged robots, typically consisting of several rigid bodies, articulated through joints
- some mobile robots can achieve locomotion on the ground without wheels or feet: e.g., snake robots



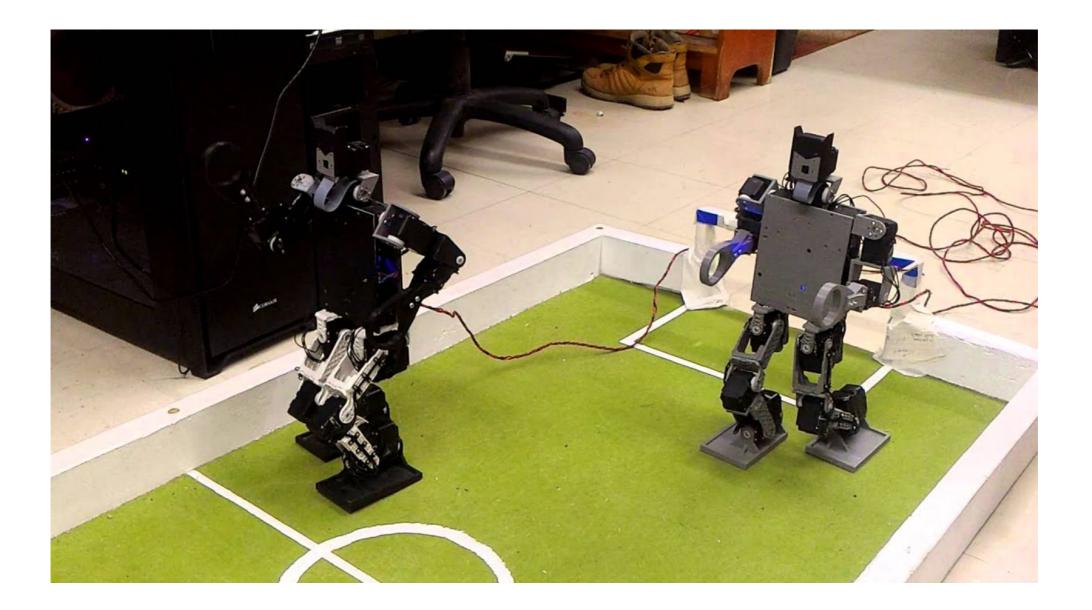
balance (not falling)

 statical balance is achieved when the projection of the robot Center of Mass (CoM) falls inside the support polygon; in the case of WMRs, one needs 3 wheels!



 dynamical balance is a different type of balance in which the CoM is replaced by the Zero Moment Point (ZMP)

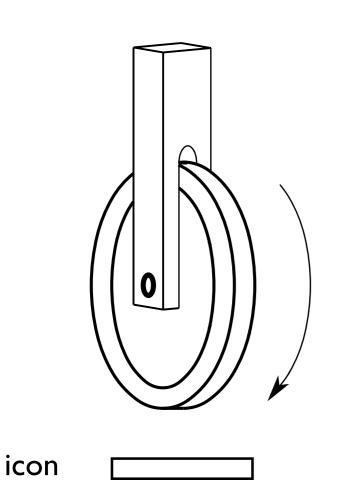
balance (not falling)



dynamic walking vs static walking

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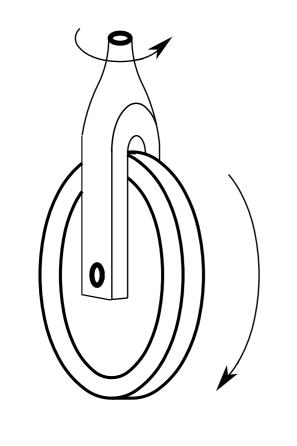
wheels: 3 basic types



fixed wheel

- fixed orientation w.r.t. the chassis
- may be active (used for driving) or passive (used for balance)

wheels: 3 basic types



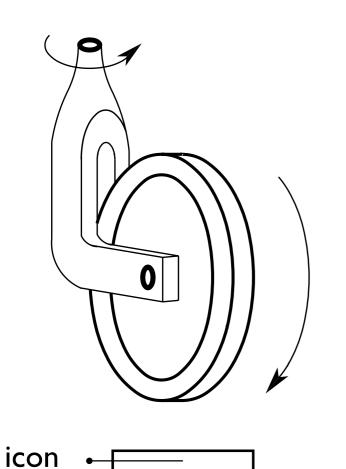
icon 🛛

orientable (steerable) wheel

• variable orientation w.r.t. the chassis

typically active (used for steering)

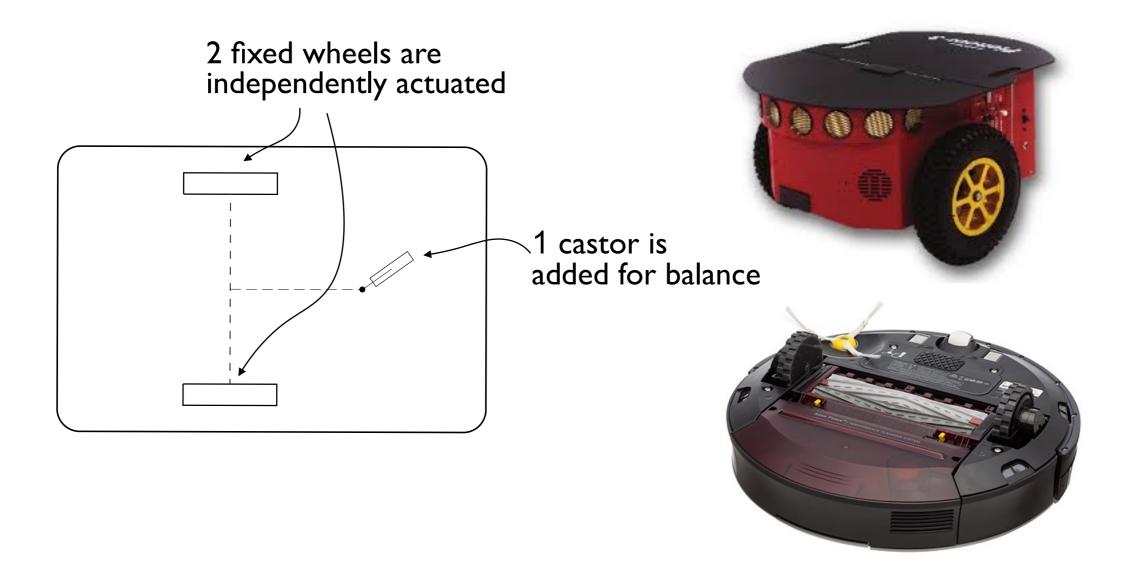
wheels: 3 basic types



caster wheel

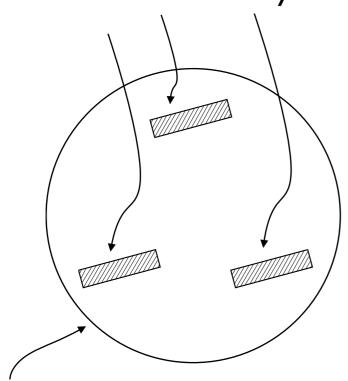
- variable orientation w.r.t. the chassis
- automatically aligns with the direction of motion





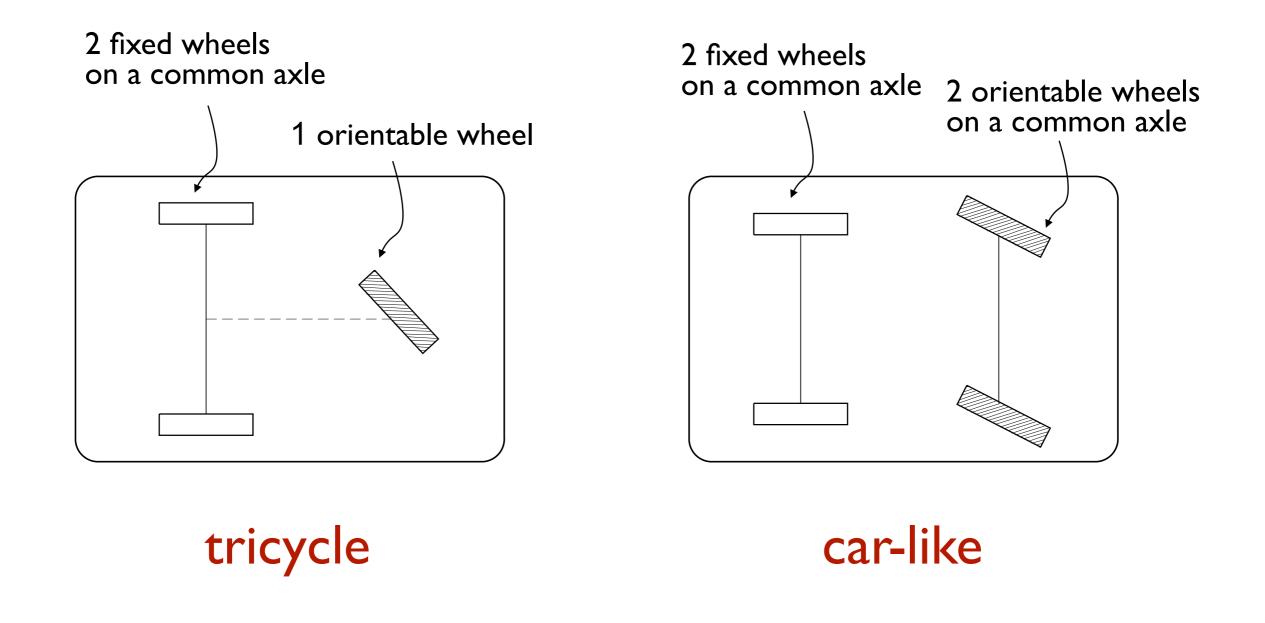
differential-drive mobile robot

3 orientable wheels are simultaneously actuated



the orientation of the chassis remains constant!

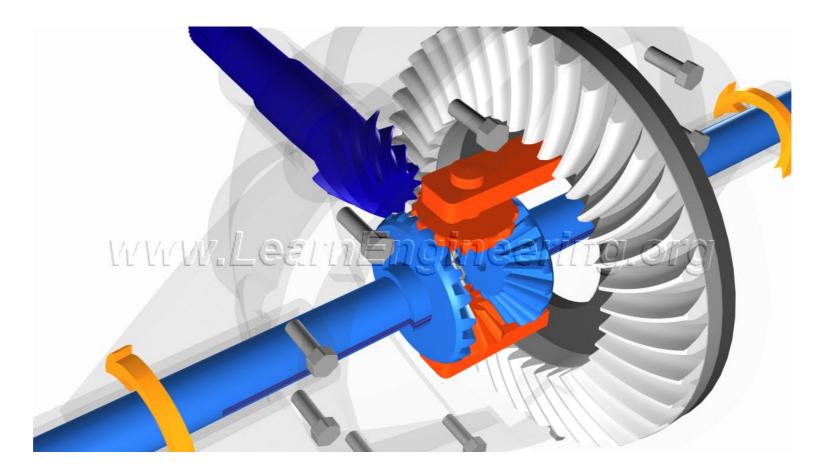
synchro-drive mobile robot



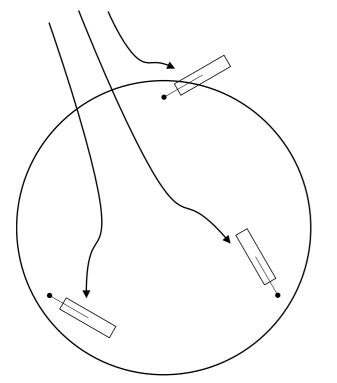
• both may be front-wheel drive or rear-wheel drive!

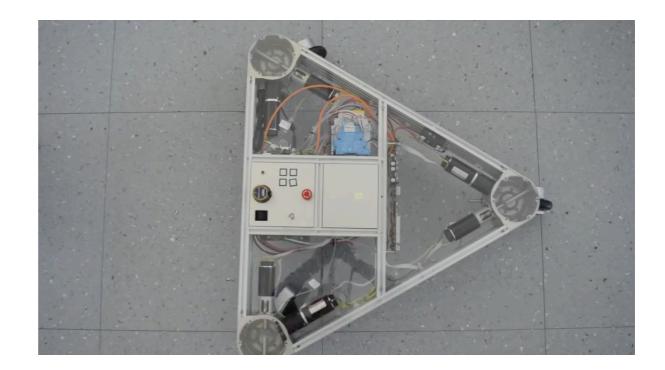
differential

- needed whenever two driving wheels are mounted on a common axle
- a mechanical device that allows the two wheels to move at different speeds

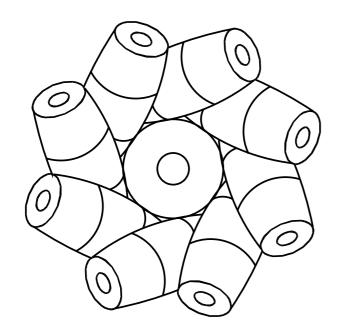


3 active castor wheels





omnidirectional mobile robot with 3 (actuated) caster wheels

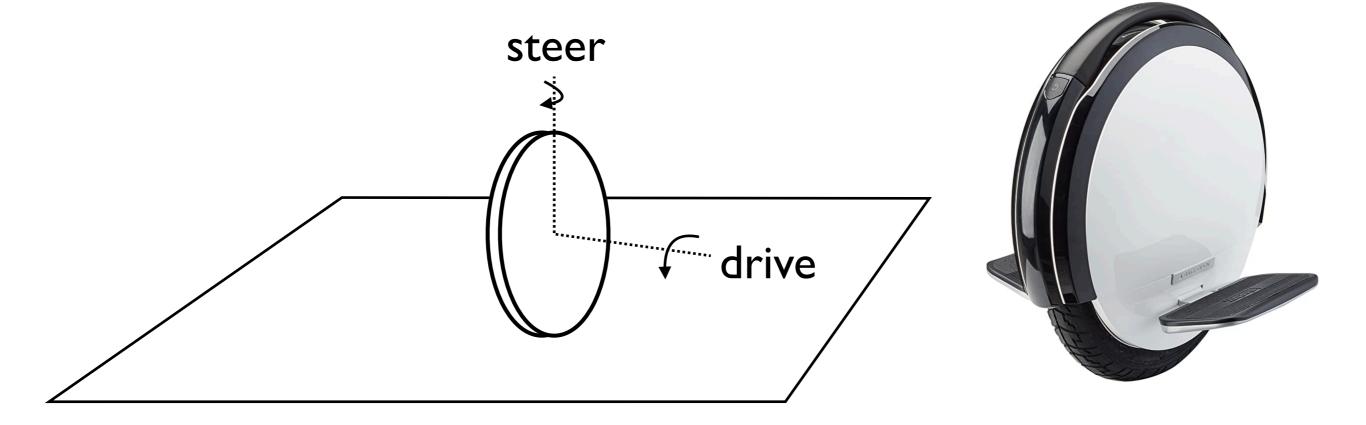


Mecanum (Swedish) wheels can be also used to build omnidirectional mobile robots

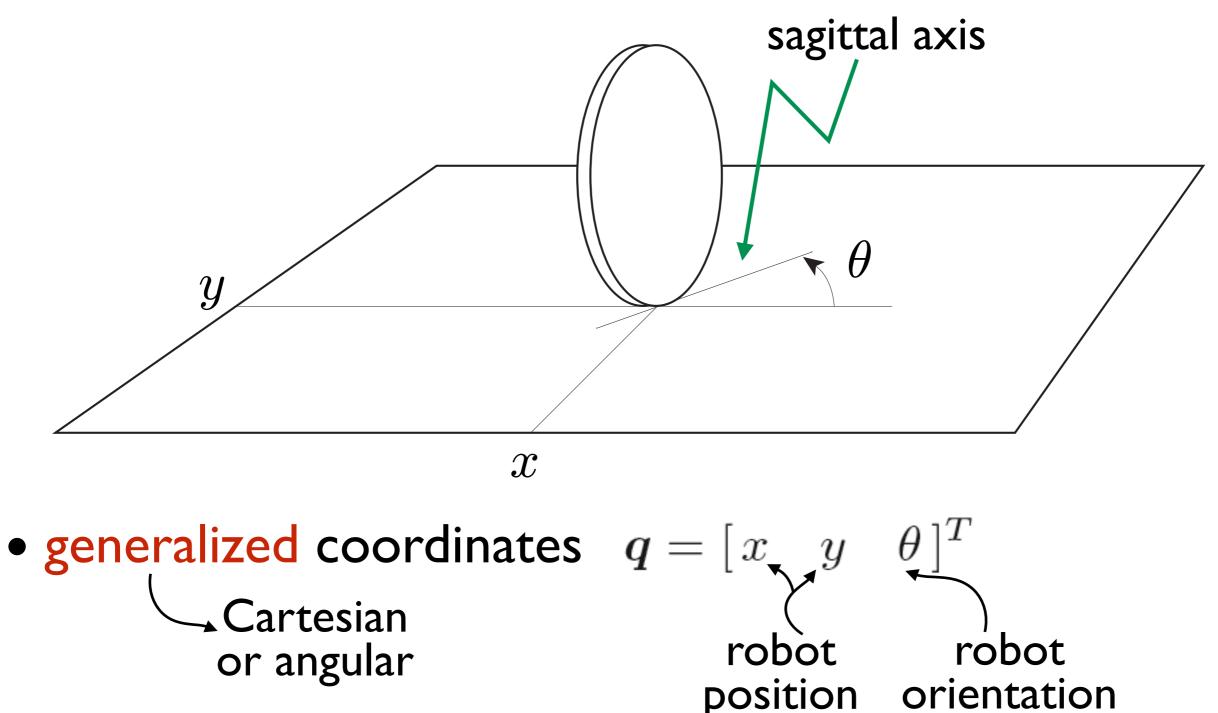


unicycle

- the simplest WMR we can conceive has a single wheel which can both drive (roll on the ground) and steer (rotate around the vertical axis)
- this rather abstract robot is called a unicycle
- a real unicycle would be unbalanced...

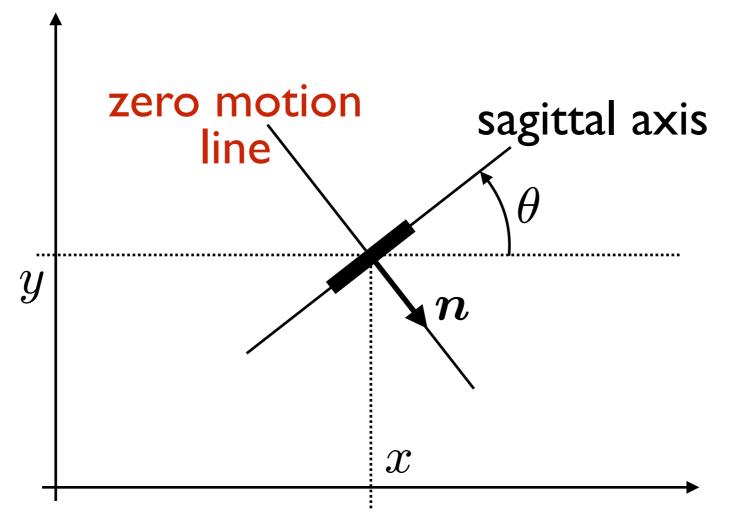


unicycle: kinematic model



• q is called configuration vector

unicycle: kinematic model



• if the wheel does not slip, we have

$$(\sin\theta - \cos\theta) \begin{pmatrix} \dot{x} \\ \dot{y} \end{pmatrix} = n^T \begin{pmatrix} \dot{x} \\ \dot{y} \end{pmatrix} = 0$$
 pure rolling constraint

no instantaneous motion along the zero motion line!

nonholonomic constraints

- \bullet the pure rolling constraint involves both generalized coordinates q and generalized velocities \dot{q}
- such constraints are called kinematic, and they may be integrable (holonomic) or not (nonholonomic)
- for example

 $x \dot{x} + y \dot{y} = 0$ can be integrated as $x^2 + y^2 = r^2$ the robot must move on a circle around the origin!

- the pure rolling constraint is instead nonholonomic
- a system with a holonomic constraint cannot reach all configurations, while a nonholonomic constraint does not prevent this

unicycle: kinematic model

- how do we derive a model from this constraint?
- rewrite the constraint so that all coordinates appear

$$(\sin\theta - \cos\theta \ 0) \begin{pmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{pmatrix} = \mathbf{A}^T(\mathbf{q}) \dot{\mathbf{q}} = 0$$

• then all admissible generalized velocities at q belong to the null space of ${m A}^T({m q})$

$$\dot{\boldsymbol{q}} = v \begin{pmatrix} \cos \theta \\ \sin \theta \\ 0 \end{pmatrix} + \omega \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \quad \begin{array}{l} \text{kinematic} \\ \text{model} \\ \end{array}$$

$$a \text{ basis of } \mathcal{N}(\boldsymbol{A}^{T}(\boldsymbol{q}))$$

unicycle: kinematic model

since

$$\dot{x}^2 + \dot{y}^2 = v^2$$
 and $\dot{\theta} = \omega$

v is the driving velocity (modulus of Cartesian velocity) and ω is the steering velocity of the robot

• the kinematic model can be rewritten as

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

a dynamical system!

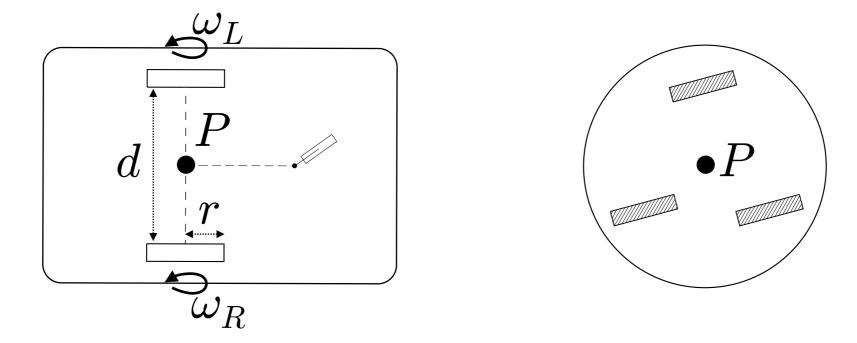
 \bullet with state ${\pmb q}$ and inputs v,ω

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- nonlinear in the state
- driftless
- by "kinematic model" here we mean "a description of all admissible instantaneous motions of the vehicle"

unicycle: equivalent vehicles

 the differential drive and the synchro drive robot are mechanically balanced versions of the unicycle

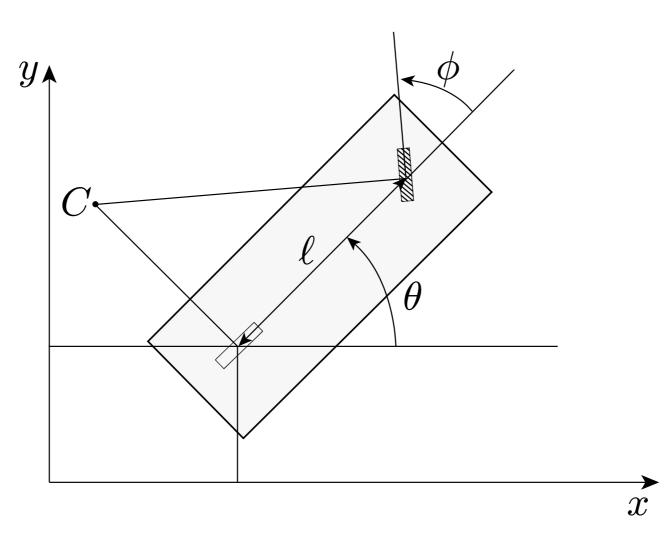


- if (x,y) denotes the position of P, their kinematic models are the same of the unicycle
- input transformation for the differential drive

$$v = \frac{r(\omega_R + \omega_L)}{2} \qquad \omega = \frac{r(\omega_R - \omega_L)}{d}$$

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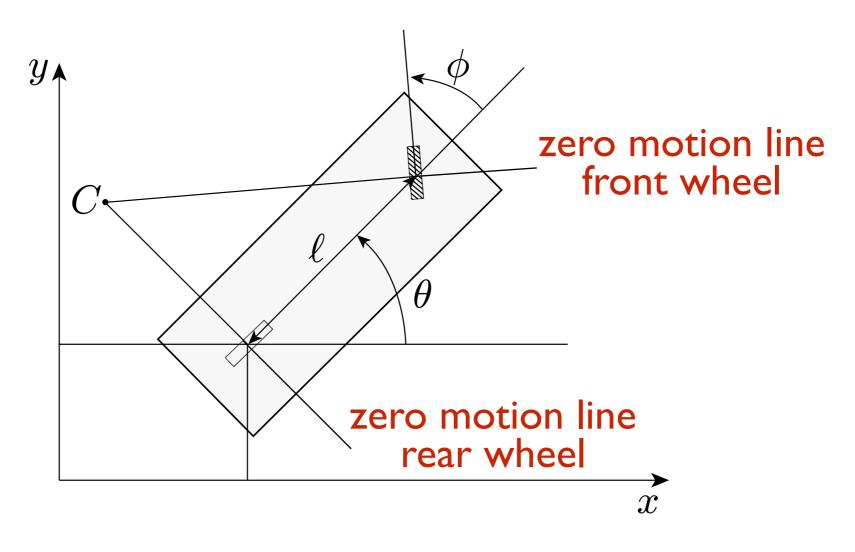
car-like robot: kinematic model



- collapse front wheels and rear wheels (bicycle model)
- generalized coordinates $q = \begin{bmatrix} x & y & \theta & \phi \end{bmatrix}^T$ robot $y & \uparrow & \uparrow$ position robot robot steering orientation angle

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car-like robot: kinematic model



- two pure rolling constraints, one for each wheel
- $\boldsymbol{A}^{T}(\boldsymbol{q})$ is then a $2{ imes}4$ matrix
- the two zero motion lines meet at a point C, called the instantaneous center of rotation (ICR)

car-like robot: kinematic model

- all admissible generalized velocities at $m{q}$ belong to the null space of $m{A}^T(m{q})$, which is now 2-dimensional
- this leads to the following kinematic model

$$\dot{\boldsymbol{q}} = v \begin{pmatrix} \cos \theta \\ \sin \theta \\ (\tan \phi)/\ell \\ 0 \end{pmatrix} + \omega \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

a basis of $\mathcal{N}(\boldsymbol{A}^{T}(\boldsymbol{q}))$

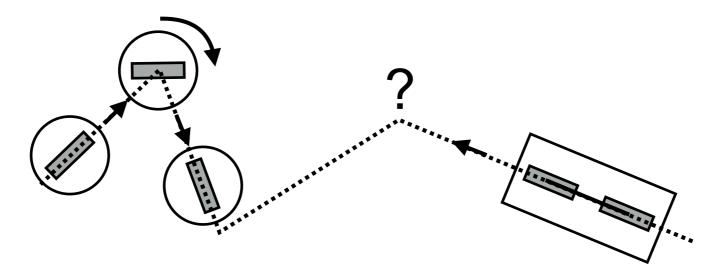
• since

$$\dot{x}^2+\dot{y}^2=v^2 \quad \text{and} \quad \dot{\phi}=\omega$$

v is the driving velocity (assuming rear-wheel drive) and ω is the steering velocity of the robot

unicycle vs car-like robot: mobility

• the unicycle can follow Cartesian paths with corners, as it can rotate on the spot; the car-like robot cannot



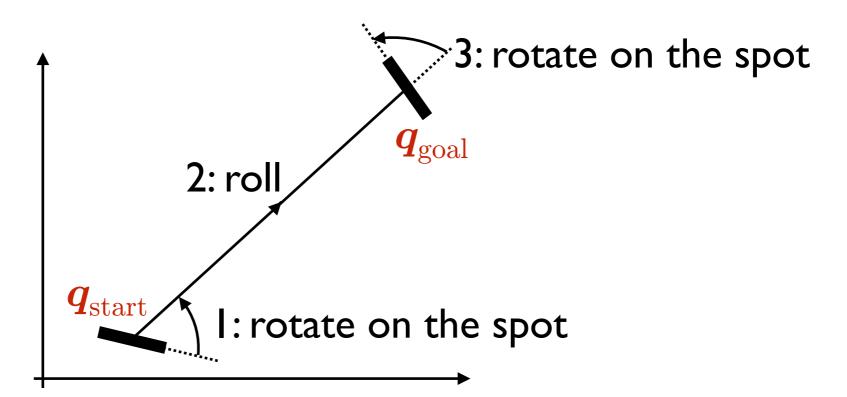
 the unicycle can handle curvature jumps; the car-like robot needs to stop and turn its steering wheel



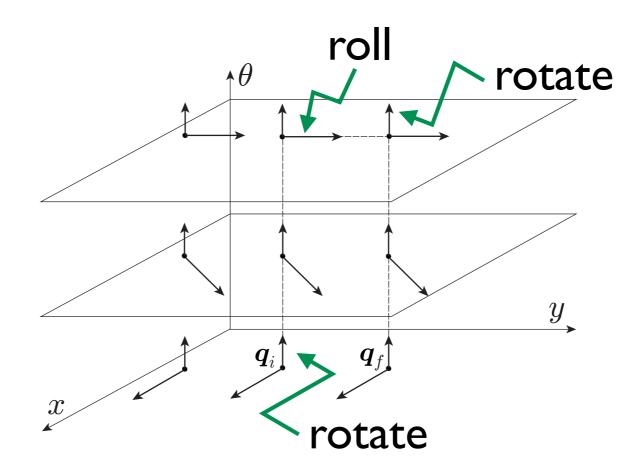
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controllability

- in spite of the different mobility, both robots are controllable: i.e., it is possible to drive from any configuration to any other configuration (parking)
- unicycle: rotate on the spot until the robot is pointing towards the final position; roll to it; rotate on the spot again to achieve the final orientation



• the same maneuver in the configuration space ${\mathcal C}$



- controllability holds also in the presence of obstacles, as long as the size of the robot allows it
- similar but more complicated maneuvers can be devised for the car-like robot

odometric localization

- localization is a procedure for estimating the robot configuration q, typically in real time (where am I?), for planning and control purposes
- \bullet in robot manipulators, joint encoders provide a direct measure of q
- WMRs are equipped with incremental encoders that measure only the rotation of the wheels, not the position and orientation of the vehicle
- a basic feature is odometric localization, in which the kinematic model is used to keep track of the motion of the motion

- consider a unicycle under constant velocity inputs v_k, ω_k in $[t_k, t_{k+1}]$, as in a digital control implementation; in each sampling interval, the robot moves along an arc of circle of radius v_k/ω_k (a line segment if $\omega_k=0$)
- assume q_k , v_k and ω_k are known; compute q_{k+1} by integration of the kinematic model over $[t_k, t_{k+1}]$
- first possibility: Euler integration

$$\begin{aligned} x_{k+1} &= x_k + v_k T_s \cos \theta_k \\ y_{k+1} &= y_k + v_k T_s \sin \theta_k \\ \theta_{k+1} &= \theta_k + \omega_k T_s \end{aligned} \qquad T_s = t_{k+1} - t_k \end{aligned}$$

• x_{k+1} and y_{k+1} are approximate; θ_{k+1} is exact

second possibility: 2nd order Runge-Kutta integration

$$x_{k+1} = x_k + v_k T_s \cos\left(\theta_k + \frac{\omega_k T_s}{2}\right)$$
$$y_{k+1} = y_k + v_k T_s \sin\left(\theta_k + \frac{\omega_k T_s}{2}\right)$$
$$\theta_{k+1} = \theta_k + \omega_k T_s$$

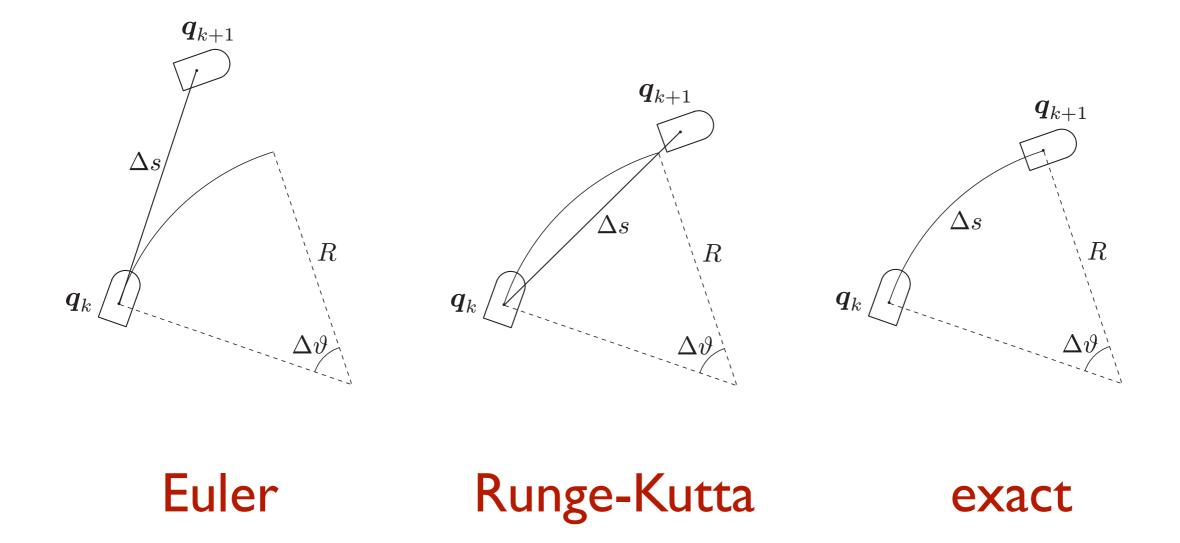
- the average orientation during $[t_k, t_{k+1}]$ is used
- as a consequence, x_{k+1} and y_{k+1} are still approximate, but more accurate

• third possibility: exact integration

$$x_{k+1} = x_k + \frac{v_k}{\omega_k} (\sin \theta_{k+1} - \sin \theta_k)$$
$$y_{k+1} = y_k - \frac{v_k}{\omega_k} (\cos \theta_{k+1} - \cos \theta_k)$$
$$\theta_{k+1} = \theta_k + \omega_k T_s$$

- for $\omega_k=0$, x_{k+1} and y_{k+1} are still defined and coincide with the solution by Euler and Runge-Kutta
- for $\omega_k \approx 0$, a conditional instruction may be used in the implementation

geometric comparison



- in practice, due to the non-ideality of any actuation system, the commanded inputs v_k and ω_k are not used
- instead, measure the effect of the actual inputs:

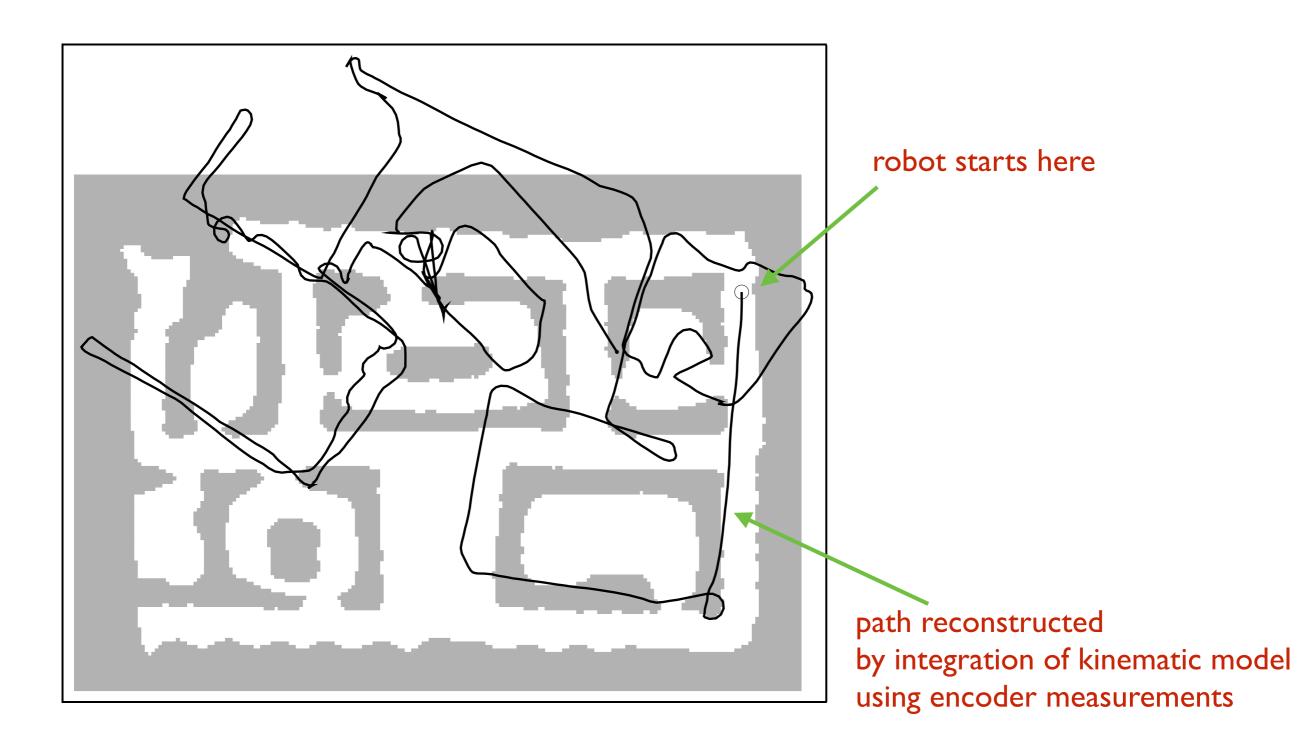
$$v_k T_s = \Delta s \qquad \omega_k T_s = \Delta \theta \qquad \frac{v_k}{\omega_k} = \frac{\Delta s}{\Delta \theta}$$

- Δs (traveled length) and $\Delta \theta$ (total orientation change) are reconstructed via proprioceptive sensors
- for example, for a differential-drive robot

$$\Delta s = \frac{r}{2} \left(\Delta \phi_R + \Delta \phi_L \right) \qquad \Delta \theta = \frac{r}{d} \left(\Delta \phi_R - \Delta \phi_L \right)$$

where $\Delta \phi_R$ and $\Delta \phi_L$ are the total rotations measured
by the wheel encoders

- odometric localization (also called dead reckoning) is subject to an error (odometric drift) that grows over time, and will typically become unacceptably large over sufficiently long paths
- causes include wheel slippage (model perturbation), inaccurate calibration of, e.g., wheel radius (model uncertainty) or numerical integration errors
- effective localization methods use proprioceptive as well as exteroceptive sensors
- the latter provide a local view of the environment which is continuously compared with what is known (the map) to correct the estimate



a typical dead reckoning result