

Kinematic and Dynamic Models

ME/CS 132b

Advanced Robotics: Navigation and Perception
3/29/2011

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Lecture Outline

Nomenclature

Kinematic Constraints and
Models

Dynamic Constraints and
Models

Applications

Current and Active
Research



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Nomenclature

State $\mathbf{x} = [\mathbf{r} \quad \mathbf{o} \quad \mathbf{v} \quad \omega \quad \dots]^T$

State Space \mathbf{X} Configuration Space \mathcal{C}

Configuration \mathbf{q}

Action \mathbf{u}

Action Space $\mathbf{U}(\mathbf{x}) \quad \mathbf{U}(\mathbf{q})$
(state-space) (configuration-space)

State/Configuration
Transition Equation $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u}) \quad \dot{\mathbf{q}} = \mathbf{f}(\mathbf{q}, \mathbf{u})$
(state-space) (configuration-space)

Trajectory $\mathbf{x}(t) = [\mathbf{x}(t_0) \quad \mathbf{x}(t_1) \quad \mathbf{x}(t_2) \quad \dots]^T$
(state-space)



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Actions

U is the set of all possible actions over all states

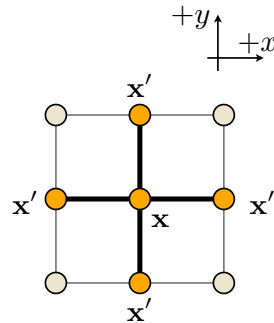
$$U = \bigcup_{\mathbf{x} \in X} U(\mathbf{x})$$

Example: Moving on a 2D grid

$$\mathbf{x} = \begin{bmatrix} x & y \end{bmatrix}^T$$

$$\mathbf{x}' = \mathbf{f}(\mathbf{x}, \mathbf{u}) = \mathbf{x} + \mathbf{u}$$

$$U = \{(0, 1), (0, -1), (1, 0), (-1, 0)\}$$



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Actions (cont.)

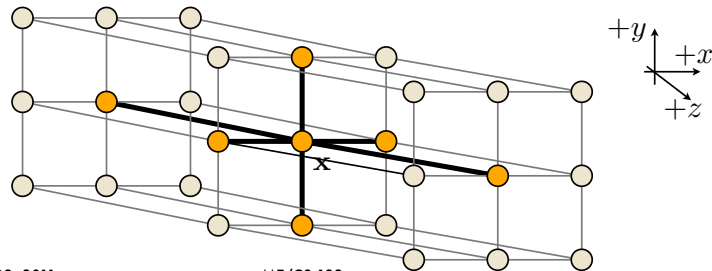
Example: Moving on a 3D grid

$$U = \bigcup_{\mathbf{x} \in X} U(\mathbf{x})$$

$$\mathbf{x} = \begin{bmatrix} x & y & z \end{bmatrix}^T$$

$$\mathbf{x}' = \mathbf{f}(\mathbf{x}, \mathbf{u}) = \mathbf{x} + \mathbf{u} \quad (\text{same as before, just a higher dimensional state space})$$

$$U = \{(0, 0, 1), (0, 0, -1), (0, 1, 0), (0, -1, 0), (1, 0, 0), (-1, 0, 0)\}$$



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Actions (cont.)

In continuous spaces the next state is determined by a integrating the state transition equation

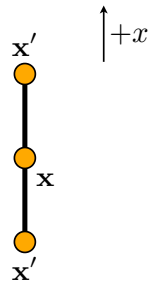
$$\mathbf{U} = \bigcup_{\mathbf{x} \in \mathbf{X}} \mathbf{U}(\mathbf{x})$$

Example: One-dimensional particle

$$\mathbf{x} = [x]^T$$

$$\dot{\mathbf{x}} = v_x$$

$$\mathbf{U} = \{v_x\} = [-1, 1]$$



Kinematic Constraints

Implicit $g(\mathbf{q}, \dot{\mathbf{q}}) \propto 0$

Parametric $\dot{\mathbf{q}} = \mathbf{f}(\mathbf{q}, \mathbf{u})$

Pfaffian $g_1(\mathbf{q})\dot{q}_1 + g_2(\mathbf{q})\dot{q}_2 + \dots + g_n(\mathbf{q})\dot{q}_n = 0$



Kinematic Constraints (cont.)

Pfaffian $g_1(\mathbf{q})\dot{q}_1 + g_2(\mathbf{q})\dot{q}_2 + \dots + g_n(\mathbf{q})\dot{q}_n = 0$

Configuration transition equation

$$\begin{aligned} \dot{q}_1 &= u_1 & \dot{q}_{n-k+1} &= f_{n-k+1}(\mathbf{q}, \mathbf{u}) \\ \dot{q}_2 &= u_2 & \dot{q}_{n-k+2} &= f_{n-k+2}(\mathbf{q}, \mathbf{u}) \\ \dot{q}_{n-k} &= u_{n-k} & \dot{q}_n &= f_n(\mathbf{q}, \mathbf{u}) \end{aligned}$$

(these are determined by solving for the remaining variables)



Kinematic Constraints (cont.)

A quick example...

$\mathcal{C} = \mathbb{R}^3$ (three dimensional state space, $n=3$)

$7\dot{q}_1 - 3\dot{q}_2 + \dot{q}_3 = 0$ (one Pfaffian constraint, $k=1$)

(two control inputs ($n - k = 2$))

$\dot{q}_1 = u_1$

$\dot{q}_1 = u_1$

$\dot{q}_2 = u_2$

$\dot{q}_2 = u_2$

(substituting for configuration rates)

$\dot{q}_3 = -7u_1 + 3u_2$

$7u_1 - 3u_2 + \dot{q}_3 = 0$

(configuration transition equation)



Simple Car Model

Rigid body that moves in a two-dimensional configuration space

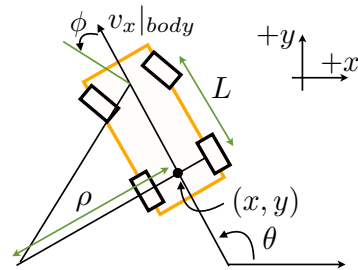
$$\mathcal{C} = \mathbb{R}^2 \times \mathbb{S}^1$$

$$q = [x \ y \ \theta]^T$$

$$\dot{x} = f_1(x, y, \theta, v_x|_{body}, \phi)$$

$$\dot{y} = f_2(x, y, \theta, v_x|_{body}, \phi)$$

$$\dot{\theta} = f_3(x, y, \theta, v_x|_{body}, \phi)$$



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Simple Car Model

Rigid body that moves in a two-dimensional configuration space

$$\mathcal{C} = \mathbb{R}^2 \times \mathbb{S}^1$$

$$q = [x \ y \ \theta]^T$$

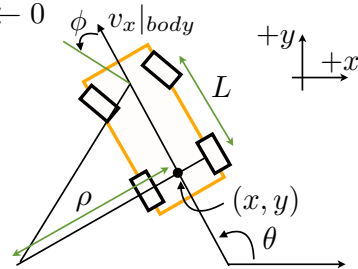
$$\frac{dy}{dx} = \tan(\theta) = \frac{\sin(\theta)}{\cos(\theta)} = \frac{\dot{y}}{\dot{x}} \text{ as } t \leftarrow 0$$

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

(rewritten as a Pfaffian constraint)

$$\dot{x} = \cos(\theta) = v_x|_{body}\cos(\theta)$$

$$\dot{y} = \sin(\theta) = v_x|_{body}\sin(\theta)$$



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Simple Car Model

How are these constraints found?

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

(n=2, k=1) \int

$$\dot{x} = u_1$$

$$\dot{y} = u_1 \tan(\theta)$$

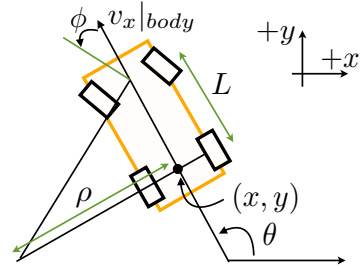
(substituting for u_1)

$$\dot{x} = \cos(\theta)$$

$$\dot{y} = \cos(\theta) \frac{\sin(\theta)}{\cos(\theta)} = \sin(\theta)$$

$$\mathcal{C} = \mathbb{R}^2 \times \mathbb{S}^1$$

$$q = [x \ y \ \theta]^T$$



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Simple Car Model

Rigid body that moves in a two-dimensional configuration space

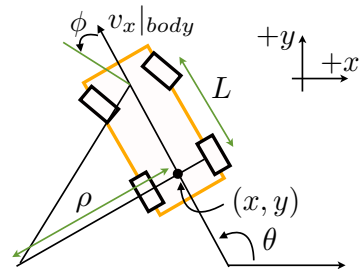
$$\mathcal{C} = \mathbb{R}^2 \times \mathbb{S}^1$$

$$q = [x \ y \ \theta]^T$$

(distance traveled) $d\omega = \rho d\theta$

$$d\theta = \frac{\tan(\phi)}{L} d\omega$$

(divide by dt) $\dot{\theta} = \frac{v_x|_{body}}{L} \tan(\phi)$



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Simple Car Model

With control inputs of the form

$$u = (u_{v_x|body}, u_\phi)$$

the configuration transition equation is

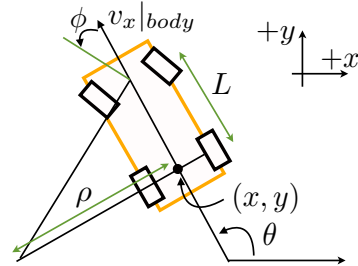
$$\dot{x} = u_{v_x|body} \cos(\theta)$$

$$\dot{y} = u_{v_x|body} \sin(\theta)$$

$$\dot{\theta} = \frac{u_{v_x|body}}{L} \tan(u_\phi)$$

$$\mathcal{C} = \mathbb{R}^2 \times \mathbb{S}^1$$

$$q = [x \ y \ \theta]^T$$



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Variations of the Simple Car Model

Simple Car

$$U = [-1, 1] \times (-\phi_{max}, \phi_{max}) \quad \text{(cannot turn in place)}$$

Tricycle

$$U = [-1, 1] \times \left(-\frac{\pi}{2}, \frac{\pi}{2}\right) \quad \text{(can turn in place)}$$

Reed-Shepp Car

$$U = \{-1, 0, 1\} \times (-\phi_{max}, \phi_{max}) \quad \text{(can drive forwards and backwards)}$$

Dubins Car

$$U = \{0, 1\} \times (-\phi_{max}, \phi_{max}) \quad \text{(cannot drive backwards)}$$



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Differential Drive Model

With inputs of the form

$$u = (u_{\omega_r}, u_{\omega_l})$$

$$\mathcal{C} = \mathbb{R}^2 \times \mathbb{S}^1$$

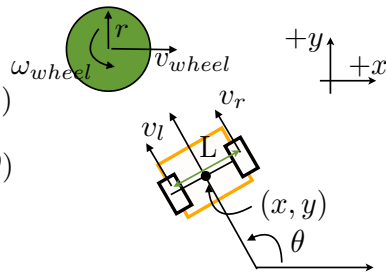
$$q = [x \ y \ \theta]^T$$

The configuration transition equation is

$$\dot{x} = \frac{r}{2}(u_{\omega_l} + u_{\omega_r})\cos(\theta)$$

$$\dot{y} = \frac{r}{2}(u_{\omega_l} + u_{\omega_r})\sin(\theta)$$

$$\dot{\theta} = \frac{r}{L}(u_{\omega_r} - u_{\omega_l})$$



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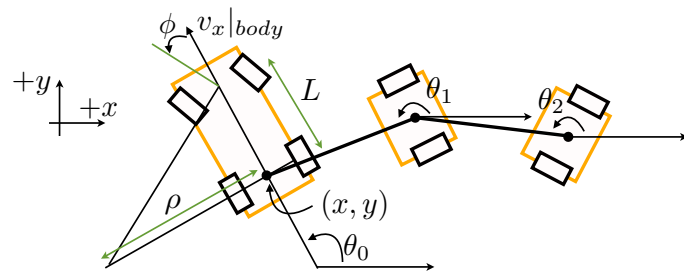
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N-Trailers

A variation of the simple car model involves one that is pulling N passively controlled trailers

$$\mathcal{C} = \mathbb{R}^2 \times \mathbb{S}^1 \times \mathbb{S}^1 \times \mathbb{S}^1$$

$$q = [x \ y \ \theta_0 \ \theta_1 \ \theta_2]^T$$

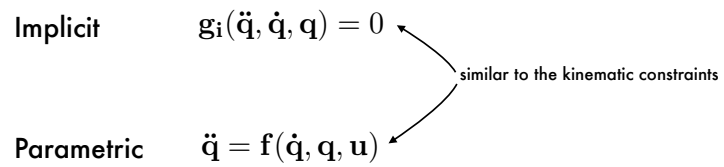


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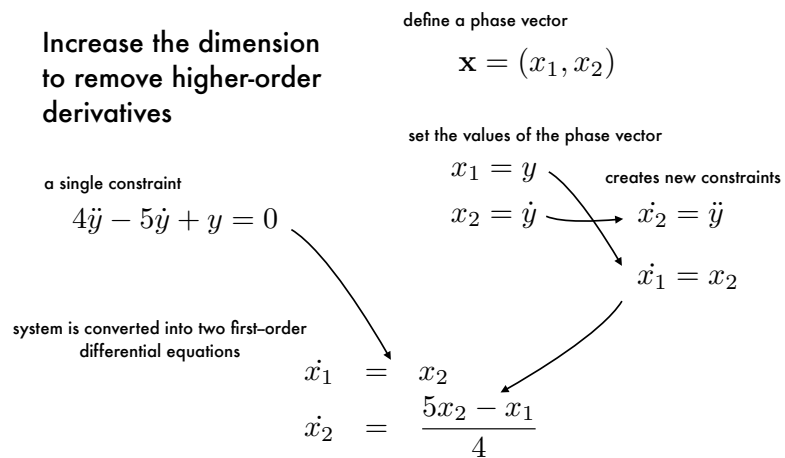
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Dynamic Constraints



Phase Space



Improved Car Model

Use a double integrator to ensure a continuously varying steering angle

$$\mathbf{x} = [x \quad y \quad \theta \quad \phi]^T$$

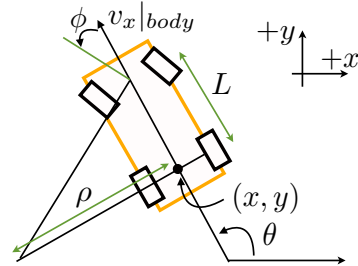
$$\mathbf{u} = (u_{v_x|body}, u_\omega)$$

$$\dot{x} = v_x|_{body} \cos(\theta)$$

$$\dot{y} = v_x|_{body} \sin(\theta)$$

$$\dot{\theta} = \frac{v_x|_{body}}{L} \tan(\phi)$$

$$\dot{\phi} = u_\omega$$



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Improved Differential Drive Model

This can similarly be applied for the differential drive model

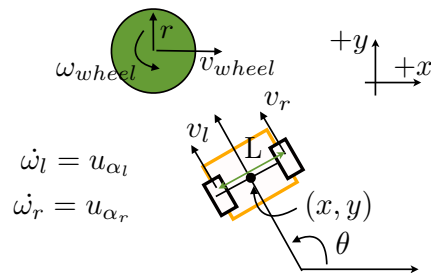
$$\mathbf{x} = [x \quad y \quad \theta \quad \omega_l \quad \omega_r]^T$$

$$\mathbf{u} = (u_{\alpha_l}, u_{\alpha_r})$$

$$\dot{x} = \frac{r}{2}(\omega_l + \omega_r) \cos(\theta)$$

$$\dot{y} = \frac{r}{2}(\omega_l + \omega_r) \sin(\theta)$$

$$\dot{\theta} = \frac{r}{L}(\omega_r - \omega_l)$$



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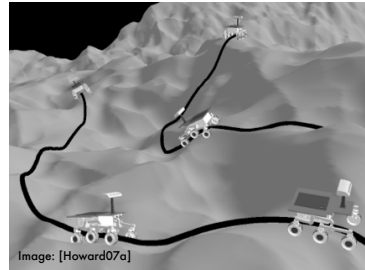
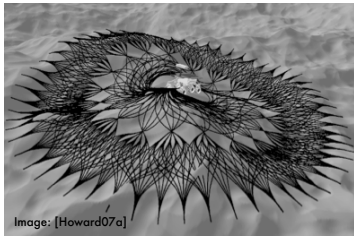
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Predictive Model Applications

Predictive Terrain Compensation [Howard07a]

Exploited a motion model that simulated the effects of terrain geometry to plan paths that considered rough terrain and wheel slip models



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Predictive Model Applications (cont.)

Autonomous automobile navigation [Ferguson08a]

Modeled curvature, curvature rate, safety
controller constraints



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Current and Active Research

Online Slip Identification [Rogers-Marcovitz10a]

Extended Kalman filter to calibrate a parameterized predictive motion model in real-time

Planar slip disturbances modeled as second-order functions of velocity and curvature

Demonstrated on multiple platforms

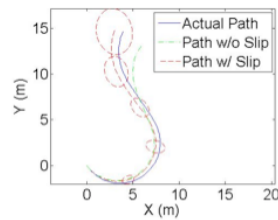


Image: [Rogers-Marcovitz10a]



Image: [Rogers-Marcovitz10a]



Next

Lecture 2: Motion simulation (3/31)

Newton-Euler Mechanics

Motion of Particles

Motion Simulation (Linear, Euler, Runge-Kutta)

Homework #1 Assigned



Document/Image References

[Ferguson08a] David Ferguson, Thomas Howard, and Maxim Likhachev, "Motion Planning in Urban Environments: Part I," Proceedings of the IEEE/RSJ 2008 International Conference on Intelligent Robots and Systems, September, 2008.

[Howard07a] Thomas Howard and Alonzo Kelly, "Optimal Rough Terrain Trajectory Generation for Wheeled Mobile Robots," International Journal of Robotics Research, Vol. 26, No. 2, March, 2007, pp. 141-166.

[Howard 09a] T.M. Howard, "Adaptive Model-Predictive Motion Planning for Navigation in Complex Environments". Ph.D. Thesis, Carnegie Mellon University, August 2009

[LaValle 06a] S. LaValle, "Planning Algorithms". Cambridge: Cambridge University Press. ISBN 0521862051.

[Rogers-Marcovitz10a] Forrest Rogers-Marcovitz and Alonzo Kelly, "On-line Mobile Robot Model Identification using Integrated Perturbative Dynamics," 12th International Symposium on Experimental Robotics, December, 2010.

