

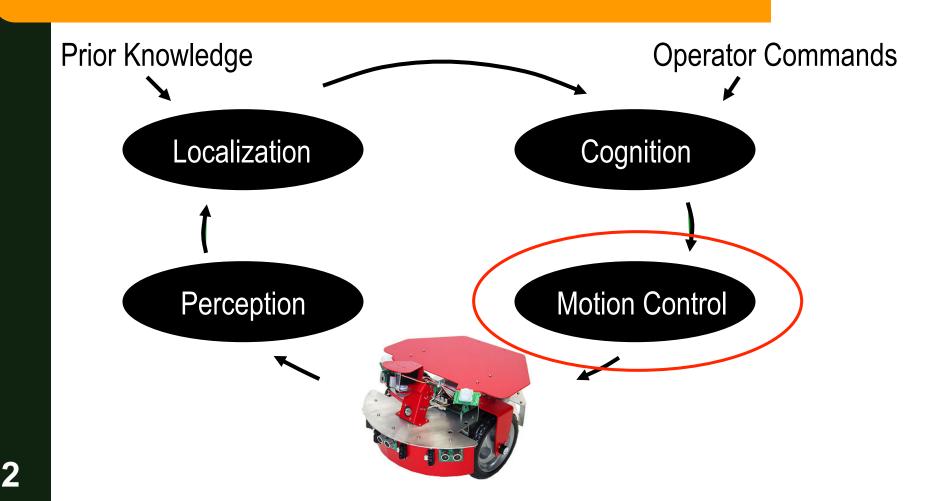
COS 495 - Lecture 6 Autonomous Robot Navigation

Instructor: Chris Clark Semester: Fall 2011

Figures courtesy of Siegwart & Nourbakhsh



Control Structure





Point Tracking

1. Linear Systems

- 2. Motion Control
- 3. Reachable Space



- Recall that the forward kinematics are a linear differential equation.
- We will use this equation to help develop a motion controller for point tracking
- We start by observing how the state x behaves if it obeys the following equation:

$$\mathbf{\dot{x}} = dx/dt = ax$$

where a is a constant



It should be obvious that the solution to the equation

$$\dot{x} = ax$$

is

$$x(t) = x_0 exp(at)$$

where

 x_0 is the initial state

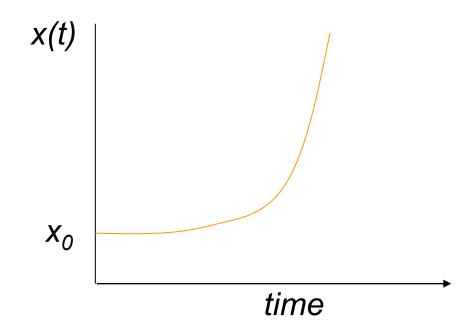


To confirm this solution, substitute into the original equation:

 $\dot{x} = ax$ $d[x_0 exp(at)]/dt = a[x_0 exp(at)]$ $ax_0 exp(at) = ax_0 exp(at)$

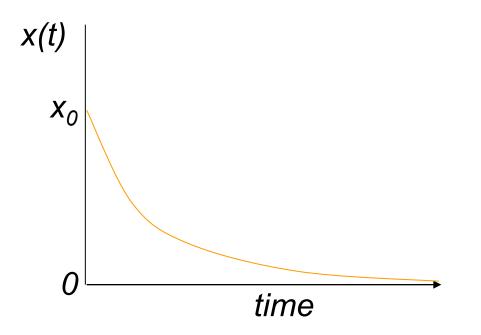


To view how the state x behaves over time, we can plot out x=x₀ exp(at), assuming a is positive:



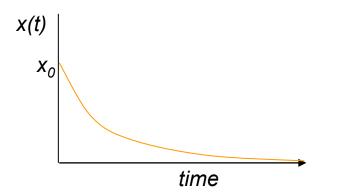


 If a is negative and we can plot out x=x₀ exp(at), we get much different results:





- This exponential decay informs us that the state x decays to zero over time.
 - We say this system is "STABLE".
 - We use this property in control theory to drive states down to zero (e.g. if $e = x_{desired} x$, drive e to 0).





- The above example was a one dimensional linear system (i.e. single state x).
- Our system is a multi-dimensional system (i.e. 3 states x, y, θ).
- We need to describe the system with matrices: $\hat{x} = Ax$

where A is a matrix such that $A \in \mathbb{R}^{n \times n}$ x is a vector such that $x \in \mathbb{R}^{1 \times n}$



In this case, the system

$$\dot{x} = Ax$$

is said to be stable if the eigen-values of A are less than 0.

The eigen values of A, represented by λ_i, are coefficients that satisfy the equation:

$$Ax_i = \lambda_i x_i$$

for particular states called x_i , called the eigen vectors.



We solve for eigen values by noting:

$$(A - \lambda I)x = 0$$

For this to hold true,

$$det (A - \lambda I) = 0$$



• Example:

$$A = \begin{bmatrix} 3 & 6\\ 1 & 4 \end{bmatrix}.$$
$$A - \lambda I = \begin{bmatrix} 3 - \lambda & 6\\ 1 & 4 - \lambda \end{bmatrix}$$
$$\det(A - \lambda I) = (3 - \lambda)(4 - \lambda) - 6$$
$$= \lambda^2 - 7\lambda + 6$$
$$= (\lambda - 6)(\lambda - 1)$$

Therefore $\lambda_1 = 6$, $\lambda_2 = 1$ The system is not stable!



- Summary:
 - If our robot behaves like a system of the form x=Ax, where the eigen values of A are negative and x represents the difference between desired and actual states, the system will move to our desired state!

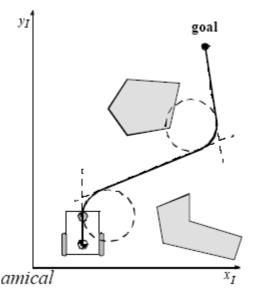


Point Tracking

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- Goal is to follow a trajectory from an initial state to some desired goal location.
- Several approaches
 - Could construct a global trajectory first, then track
 points on the trajectory locally



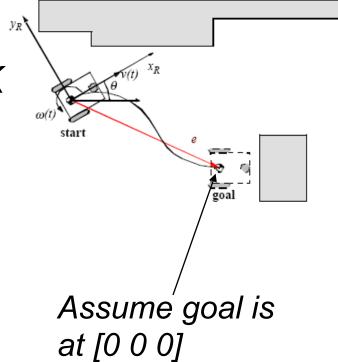


If we define the error to be in the robot frame:

 $e(t) = [x y \theta]^{T}$

Goal is to find gain matrix K such that control of v(t) and w(t) will drive the error e(t) to zero.

$$\begin{bmatrix} v(t) \\ w(t) \end{bmatrix} = Ke(t)$$



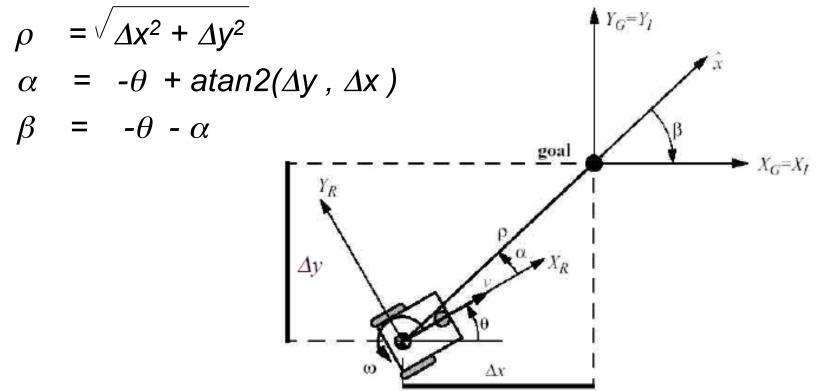


Recall our forward kinematics

$$\begin{bmatrix} \mathbf{\dot{x}} \\ \mathbf{\dot{y}} \\ \mathbf{\dot{\theta}} \end{bmatrix} = \begin{bmatrix} \cos\theta & 0 \\ \sin\theta & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{v} \\ \mathbf{w} \end{bmatrix}$$



We use the coordinate transformation





• Now we define the problem as driving the robot to goal $\begin{pmatrix} \rho \\ \alpha \\ \beta \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$



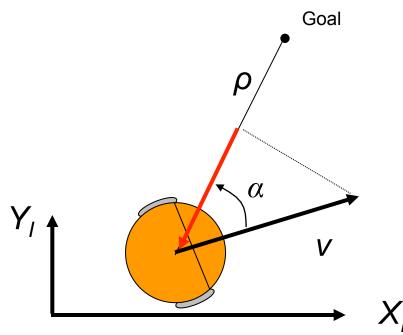
• We know this will happen if the dynamics of the system obey $\begin{bmatrix}
\dot{\rho} \\
\dot{\alpha} \\
\dot{\beta}
\end{bmatrix} = A \begin{bmatrix}
\rho \\
\alpha \\
\beta
\end{bmatrix}$

Where A is a 3x3 matrix with eigen values less than 0.



 Using the coordinate transformation, calculate the new kinematics:

$$\stackrel{\bullet}{\rho} = projection of v on \rho$$
$$= -v \cos(\alpha)$$





Goal

α

V

Motion Control

 Using the coordinate transformation, calculate the new kinematics:

$$\rho \hat{\beta} = \text{projection of v perpendicular to } \rho$$
$$= -v \sin(\alpha)$$
$$\hat{\beta} = -v \sin(\alpha) / \rho$$



 Using the coordinate transformation, calculate the new kinematics:

$$\alpha = -\beta - \theta$$

$$\dot{\alpha} = -\dot{\beta} - \dot{\theta}$$

$$\dot{\alpha} = v \sin(\alpha)/\rho - w$$



In matrix from:

$$\begin{bmatrix} \boldsymbol{\rho} \\ \boldsymbol{\alpha} \\ \boldsymbol{\beta} \\ \boldsymbol{\beta} \end{bmatrix} = \begin{bmatrix} -\cos\alpha & 0 \\ \sin\alpha / \rho & -1 \\ -\sin\alpha / \rho & 0 \end{bmatrix} \begin{bmatrix} \mathbf{v} \\ \mathbf{w} \end{bmatrix}$$

for α within (- $\pi/2$, $\pi/2$]



• Let's try the control law: $v = k_{\rho}\rho$ $w = k_{\alpha}\alpha + k_{\beta}\beta$



- To analyze controller, substitute control law into kinematics and linearize:
 - For small x, cosx \approx 1 and sinx \approx x $\begin{bmatrix}
 \bullet \\
 \rho \\
 \bullet \\
 \alpha \\
 \bullet \\
 \beta
 \end{bmatrix} = \begin{bmatrix}
 -k_{\rho} & 0 & 0 \\
 0 & -(k_{\alpha} - k_{\rho}) & -k_{\beta} \\
 0 & -k_{\rho} & 0
 \end{bmatrix} \begin{bmatrix}
 \rho \\
 \alpha \\
 \beta
 \end{bmatrix}$
 - This is in the form...



- Check for stability:
 - Take the determinant of A and solving for eigen values leads to:

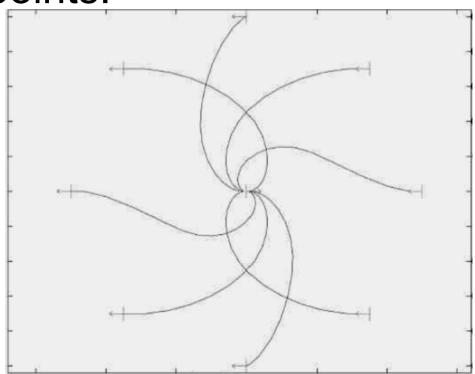
$$(\lambda + k_{\rho}) (\lambda^2 + \lambda(k_{\alpha} - k_{\rho}) - k_{\rho}k_{\beta}) = 0$$

• Thus the system will be stable if:

$$k_{\rho} > 0 \qquad k_{\beta} < 0 \qquad k_{\alpha} - k_{\rho} > 0$$



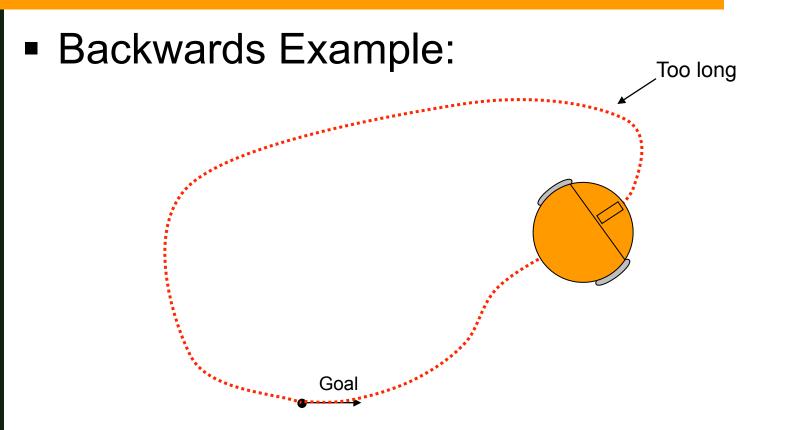
 Testing this control law with many different start points:





- The derived control law works well if $\alpha \in [-\pi/2, \pi/2]$
- For other cases where abs(α) > π/2, we must modify the controller. So that the robot will move backwards to the desired position when required





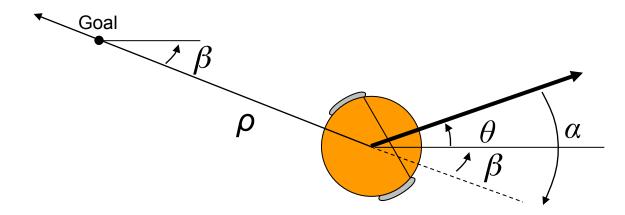


Backwards Method:

$$\rho = \sqrt{\Delta x^2 + \Delta y^2}$$

$$\alpha = -\theta + atan2(-\Delta y, -\Delta x)$$

$$\beta = -\theta - \alpha$$





- Backwards Method Summary:
 - If α ∈ [-π/2, π/2]
 - Use regular transform to polar coordinates
 - Use control law: $v = k_{\rho}\rho$ $w = k_{\alpha}\alpha + k_{\beta}\beta$
 - Else
 - Redefine α as shown in backwards method
 - Use control law: $v = -k_{\rho}\rho$ $w = k_{\alpha}\alpha + k_{\beta}\beta$



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Kinematic Constraints

 One can calculate constraints on each individual wheel, then combine for constraints on entire robot.

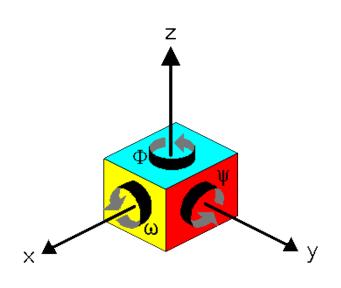
Two main constraints:

- Rolling Constraint: no slipping!
- Sliding Constraint: no lateral movement!



Degrees of Freedom:

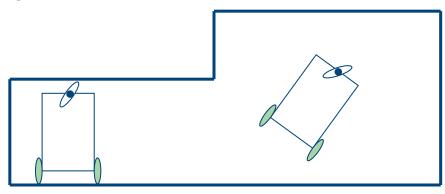
- Def' n: The number of coordinates that it takes to uniquely specify the state of a system.
- In 3D, there are 6 degrees of freedom associated to the movement of a rigid body: 3 for its position, and 3 for its orientation.



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- Configurations in the Workspace
 - A robot's workspace is defined by the Degrees Of Freedom of the robot state.
 - Not all robot configurations within the workspace are reachable



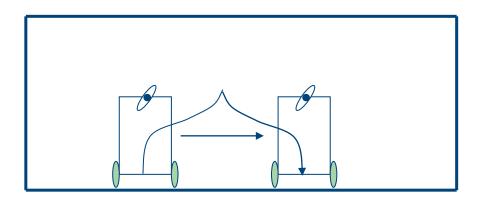


Holonomic Robots

- A robot is holonomic if it has zero nonholonomic constraints.
- A nonholonomic constraint is one that is not integrable.



- Paths in the Workspace
 - Path's in the workspace are limited, especially if the robot is nonholonomic





- Trajectories in the Workspace
 - A trajectory is a path parameterized by time.
 - Admissible paths don't always lead to admissible trajectories.

