

## Velocity Motion Model (cont)

# Velocity Motion Model

## ► center of circle

$$\begin{pmatrix} x^* \\ y^* \end{pmatrix} = \begin{pmatrix} x \\ y \end{pmatrix} + \begin{pmatrix} -\lambda \sin \theta \\ \lambda \cos \theta \end{pmatrix} = \begin{pmatrix} \frac{x+x'}{2} + \mu(y-y') \\ \frac{y+y'}{2} + \mu(x'-x) \end{pmatrix}$$

where

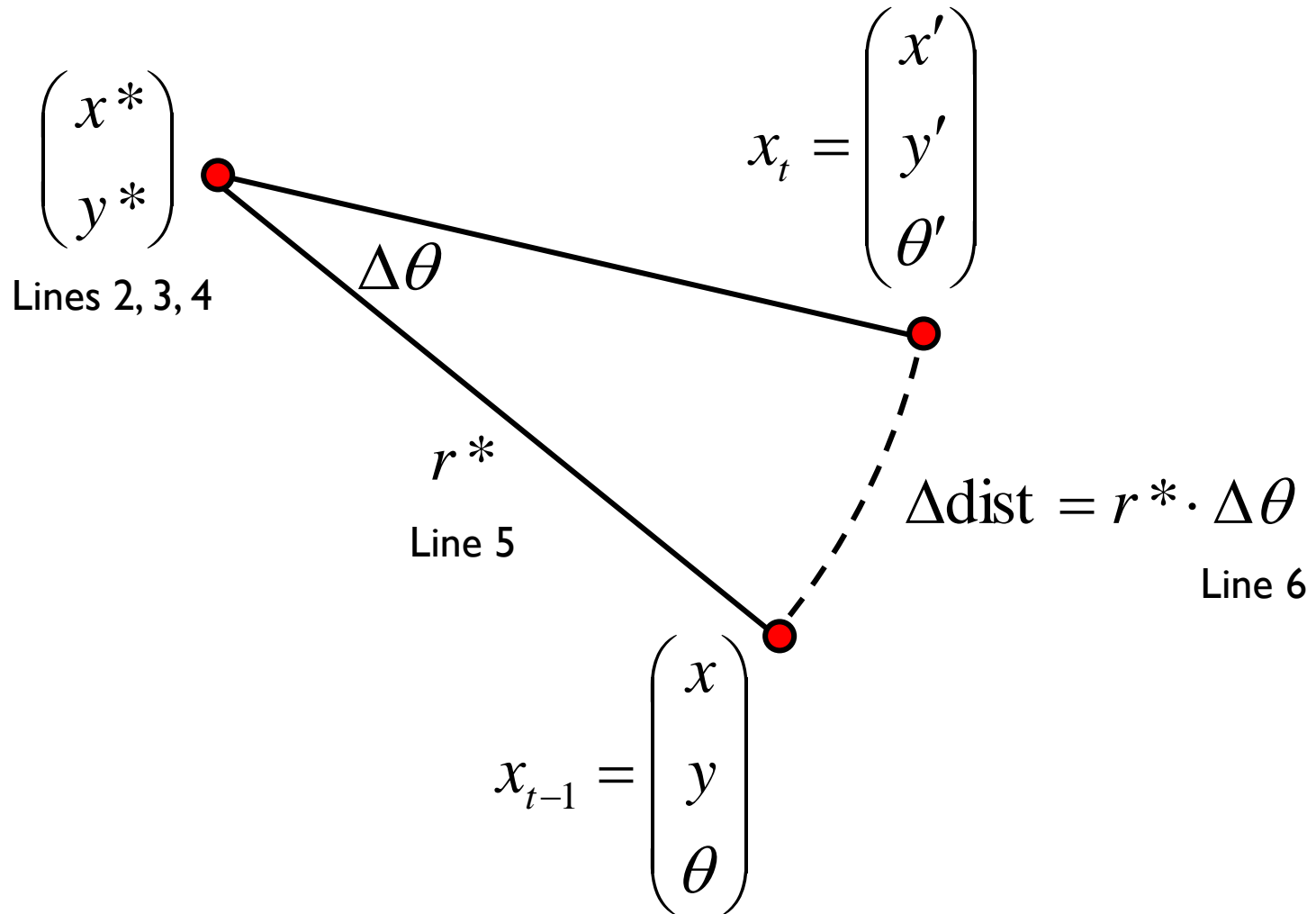
$$\mu = \frac{1}{2} \frac{(x-x') \cos \theta + (y-y') \sin \theta}{(y-y') \cos \theta - (x-x') \sin \theta}$$

# Velocity Motion Model

```
1:  Algorithm motion_model_velocity( $x_t, u_t, x_{t-1}$ ):
2:       $\mu = \frac{1}{2} \frac{(x - x') \cos \theta + (y - y') \sin \theta}{(y - y') \cos \theta - (x - x') \sin \theta}$ 
3:       $x^* = \frac{x + x'}{2} + \mu(y - y')$ 
4:       $y^* = \frac{y + y'}{2} + \mu(x' - x)$ 
5:       $r^* = \sqrt{(x - x^*)^2 + (y - y^*)^2}$ 
6:       $\Delta\theta = \text{atan2}(y' - y^*, x' - x^*) - \text{atan2}(y - y^*, x - x^*)$ 
7:       $\hat{v} = \frac{\Delta\theta}{\Delta t} r^*$ 
8:       $\hat{\omega} = \frac{\Delta\theta}{\Delta t}$ 
9:       $\hat{\gamma} = \frac{\theta' - \theta}{\Delta t} - \hat{\omega}$ 
10:     return  $\text{prob}(v - \hat{v}, \alpha_1 v^2 + \alpha_2 \omega^2) \cdot \text{prob}(\omega - \hat{\omega}, \alpha_3 v^2 + \alpha_4 \omega^2)$ 
         $\cdot \text{prob}(\hat{\gamma}, \alpha_5 v^2 + \alpha_6 \omega^2)$ 
```

# Velocity Motion Model

- ▶ rotation of  $\Delta\theta$  about  $(x^*, y^*)$  from  $(x, y)$  to  $(x', y')$  in time  $\Delta t$



# Velocity Motion Model

- ▶ given  $\Delta\theta$  and  $\Delta\text{dist}$  we can compute the velocities needed to generate the motion

$$\hat{u}_t = \begin{pmatrix} \hat{v}_t \\ \hat{\omega}_t \end{pmatrix} = \begin{pmatrix} \Delta\text{dist} / \Delta t \\ \Delta\theta / \Delta t \end{pmatrix} \quad \text{Steps 7, 8}$$

- ▶ notice what the algorithm has done
  - ▶ it has used an inverse motion model to compute the control vector that would be needed to produce the motion from  $x_{t-1}$  to  $x_t$
  - ▶ in general, the computed control vector will be different from the actual control vector  $u_t$

# Velocity Motion Model

- ▶ recall that we want the posterior conditional density

$$p(x_t | u_t, x_{t-1})$$

of the control action  $u_t$  carrying the robot from pose  $x_{t-1}$  to  $x_t$  in time  $\Delta t$

- ▶ so far the algorithm has computed the required control action  $\hat{u}_t$  needed to carry the robot from position (x y) to position (x' y')
- ▶ the control action has been computed assuming the robot moves on a circular arc

# Velocity Motion Model

▶ the computed heading of the robot is  $\hat{\theta} = \theta + \Delta\theta$

▶ the heading should be  $\theta'$

▶ the difference is  $\theta_{\text{err}} = \theta' - \hat{\theta}$   
 $= \theta' - \theta - \Delta\theta$

▶ or expressed as an angular velocity

the in-place bearing error; i.e., the amount the robot must rotate by at the final location to achieve a bearing of  $\theta'$

$$\gamma_{\text{err}} = \frac{\theta_{\text{err}}}{\Delta t}$$
$$= \frac{\theta' - \theta}{\Delta t} - \hat{\omega}$$

Line 9,  
Eq 5.25, 5.28

# Velocity Motion Model

- ▶ similarly, we can compute the errors of the computed linear and rotational velocities

$$\begin{aligned} v_{\text{err}} &= v - \hat{v} \\ &= \frac{\Delta \text{dist}}{\Delta t} \end{aligned}$$

$$\begin{aligned} \omega_{\text{err}} &= \omega - \hat{\omega} \\ &= \frac{\Delta \theta}{\Delta t} \end{aligned}$$



# Velocity Motion Model

- ▶ if we assume that the robot has independent control over its controlled linear and angular velocities then the joint density of the errors is

$$p(v_{\text{err}}, \omega_{\text{err}}, \gamma_{\text{err}}) = p(v_{\text{err}}) p(\omega_{\text{err}}) p(\gamma_{\text{err}})$$

- ▶ what do the individual densities look like?

# Velocity Motion Model

- ▶ the most common noise model is additive zero-mean noise, i.e.

$$\begin{pmatrix} \hat{v} \\ \hat{\omega} \end{pmatrix} = \begin{pmatrix} v \\ \omega \end{pmatrix} + \begin{pmatrix} v_{\text{noise}} \\ \omega_{\text{noise}} \end{pmatrix}$$

actual      commanded      noise  
velocity      velocity

- ▶ we need to decide on other characteristics of the noises
  - ▶ “spread”      variance
  - ▶ “skew”      skew
  - ▶ “peakedness”      kurtosis
- ▶ typically, only the variance is specified
  - ▶ the true variance is typically unknown

# Velocity Motion Model

- ▶ assumes that the variances can be modeled as

$$\begin{aligned}\text{var}(v_{\text{noise}}) &= \alpha_1 v^2 + \alpha_2 \omega^2 \\ \text{var}(\omega_{\text{noise}}) &= \alpha_3 v^2 + \alpha_4 \omega^2\end{aligned}\quad \text{Eq 5.10}$$

where the  $\alpha_i$  are robot specific error parameters

- ▶ the less accurate the robot the larger the  $\alpha_i$
- ▶ assumption models standard deviation of the noise is proportional to the commanded velocity of the robot

# Velocity Motion Model

- ▶ a robot travelling on a circular arc has no independent control over its heading
  - ▶ the heading must be tangent to the arc

$$\theta' = \theta + \hat{\omega} \Delta t$$

- ▶ this is problematic if you have a noisy commanded angular velocity  $\omega$
- ▶ thus, we assume that the final heading is actually given by

$$\theta' = \theta + \hat{\omega} \Delta t + \hat{\gamma} \Delta t \quad \text{Eq 5.14}$$

where  $\hat{\gamma}$  is the angular velocity of the robot spinning in place

# Velocity Motion Model

- assumes that

$$\hat{\gamma} = \underset{\text{actual velocity}}{0} + \underset{\text{noise}}{\gamma_{\text{noise}}}$$

where

$$\text{var}(\gamma_{\text{noise}}) = \alpha_5 v^2 + \alpha_6 \omega^2 \quad \text{Eq 5.15}$$