

Joint Probability

- ▶ example: insurance policy deductibles

		y			
		\$0	\$100	\$200	← home
x	\$100	0.20	0.10	0.20	
	\$250	0.05	0.15	0.30	

↑
automobile

Joint Probability and Independence

- ▶ X and Y are said to be independent if

$$P(x,y) = P(x) P(y)$$

for all possible values of x and y

- ▶ example: two fair dice

$$P(X=\text{even and } Y=\text{even}) = (1/2) (1/2)$$

$$P(X=1 \text{ and } Y=\text{not } 1) = (1/6) (5/6)$$

- ▶ are X and Y independent in the insurance deductible example?

Marginal Probabilities

- ▶ the marginal probability distribution of X

$$P_X(x) = \sum_y P(x, y)$$

describes the probability of the event that X has the value x

- ▶ similarly, the marginal probability distribution of Y

$$P_Y(y) = \sum_x P(x, y)$$

describes the probability of the event that Y has the value y

Joint Probability

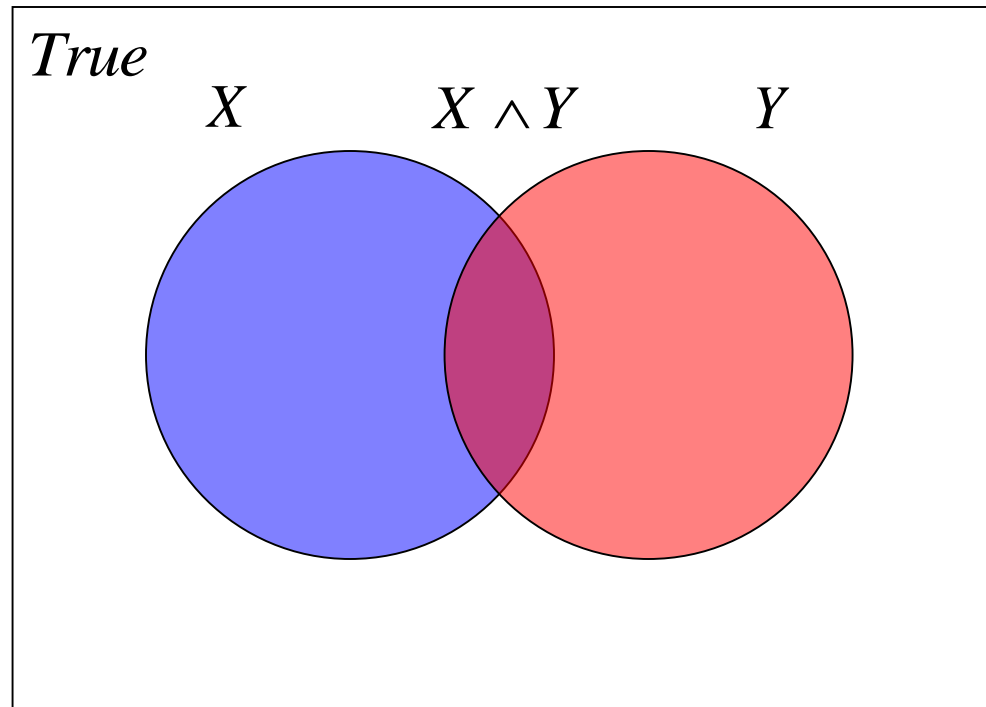
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Conditional Probability

- ▶ the conditional probability $P(x | y) = P(X=x | Y=y)$ is the probability of $P(X=x)$ if $Y=y$ is known to be true
 - ▶ “conditional probability of x given y ”



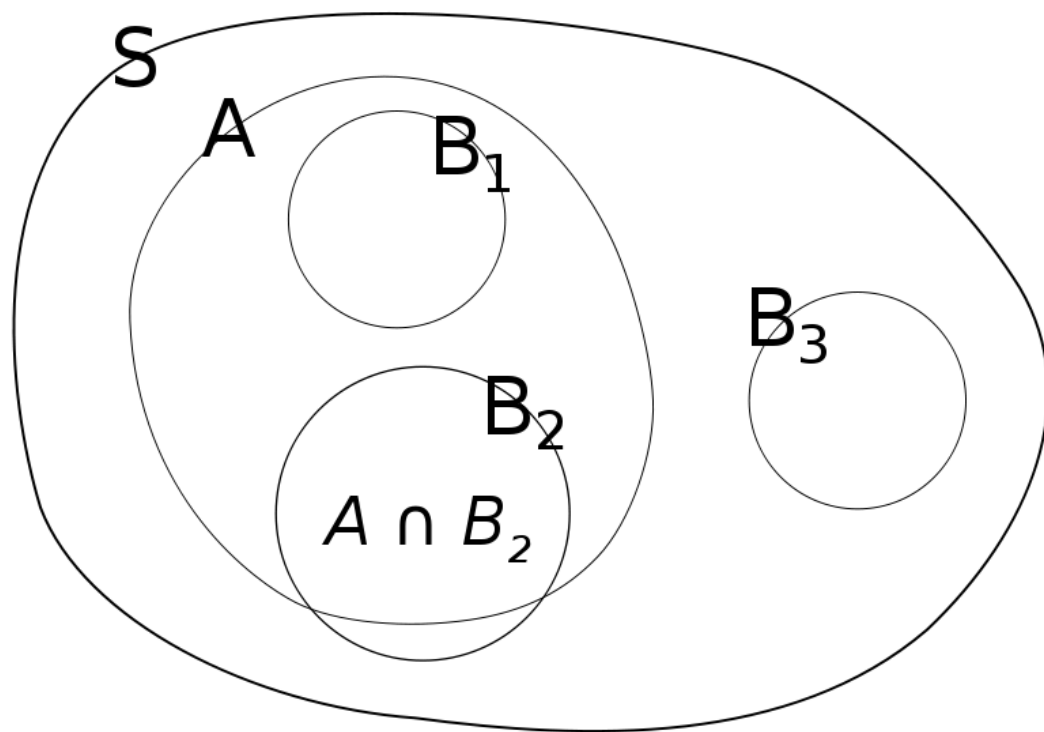
Conditional Probability

$$P(A) \approx 0.3$$

$$P(A | B_3) = ?$$

$$P(A | B_1) = ?$$

$$P(A | B_2) = ?$$



Conditional Probability

- ▶ “information changes probabilities”
- ▶ example:
 - ▶ roll a fair die; what is the probability that the number is a 3?
 - ▶ what is the probability that the number is a 3 if someone tells you that the number is odd? is even?
- ▶ example:
 - ▶ pick a playing card from a standard deck; what is the probability that it is the ace of hearts?
 - ▶ what is the probability that it is the ace of hearts if someone tells you that it is an ace? that it is a heart? that it is a king?

Conditional Probability

$$P(x | y) = \frac{P(x, y)}{P(y)}$$

- ▶ if X and Y are independent then

$$P(x, y) = P(x)P(y)$$

$$\therefore P(x | y) = \frac{P(x)P(y)}{P(y)} = P(x)$$

Bayes Formula

$$P(x, y) = P(x | y)P(y) = P(y | x)P(x)$$

\Rightarrow

$$\underbrace{P(x | y)}_{\text{posterior}} = \frac{P(y | x) P(x)}{P(y)} = \frac{\text{likelihood} \cdot \text{prior}}{\text{evidence}}$$

Bayes Rule with Background Knowledge

$$P(x \mid y, z) = \frac{P(y \mid x, z) P(x \mid z)}{P(y \mid z)}$$

Back to Kinematics

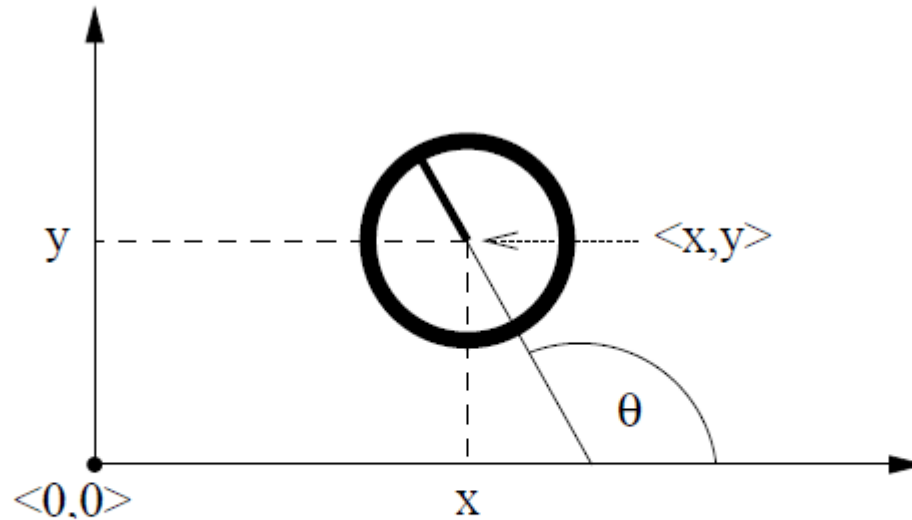


Figure 5.1 Robot pose, shown in a global coordinate system.

pose vector or state $x_t = \begin{bmatrix} x \\ y \\ \theta \end{bmatrix}$ $\left. \begin{array}{l} \text{location (in world frame)} \\ \text{bearing or heading} \end{array} \right\}$

Probabilistic Robotics

- ▶ we seek the conditional density

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

- ▶ what is the density of the state

$$x_t$$

given the motion command

$$u_t$$

performed at

$$x_{t-1}$$

Probabilistic Robotics

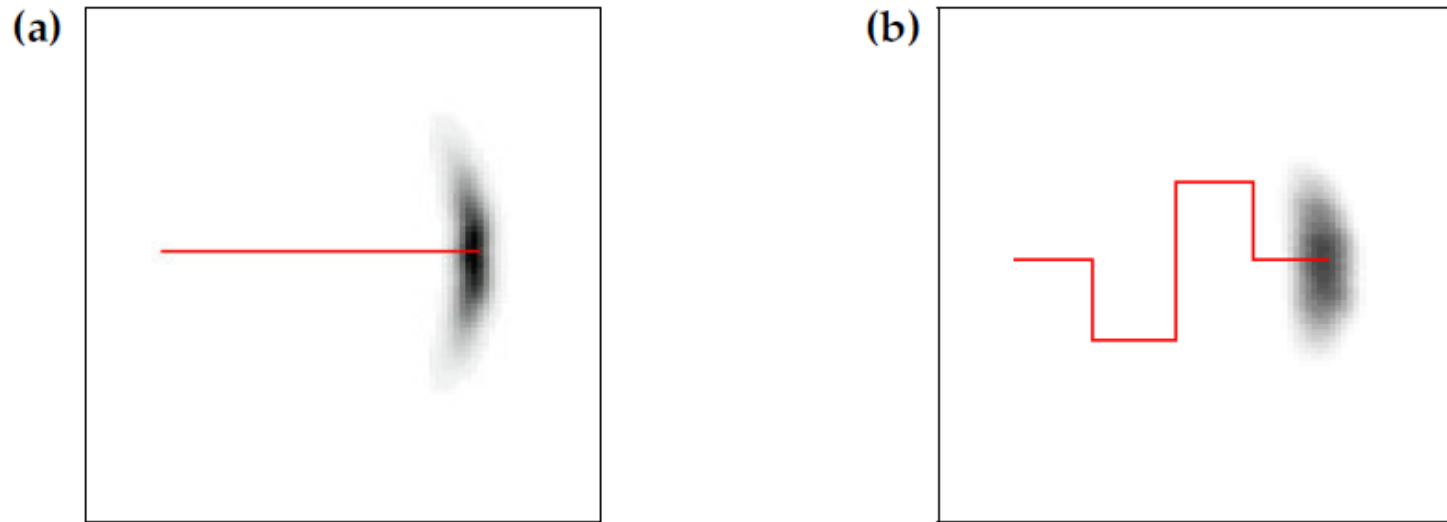


Figure 5.2 The motion model: Posterior distributions of the robot's pose upon executing the motion command illustrated by the solid line. The darker a location, the more likely it is. This plot has been projected into 2-D. The original density is three-dimensional, taking the robot's heading direction θ into account.

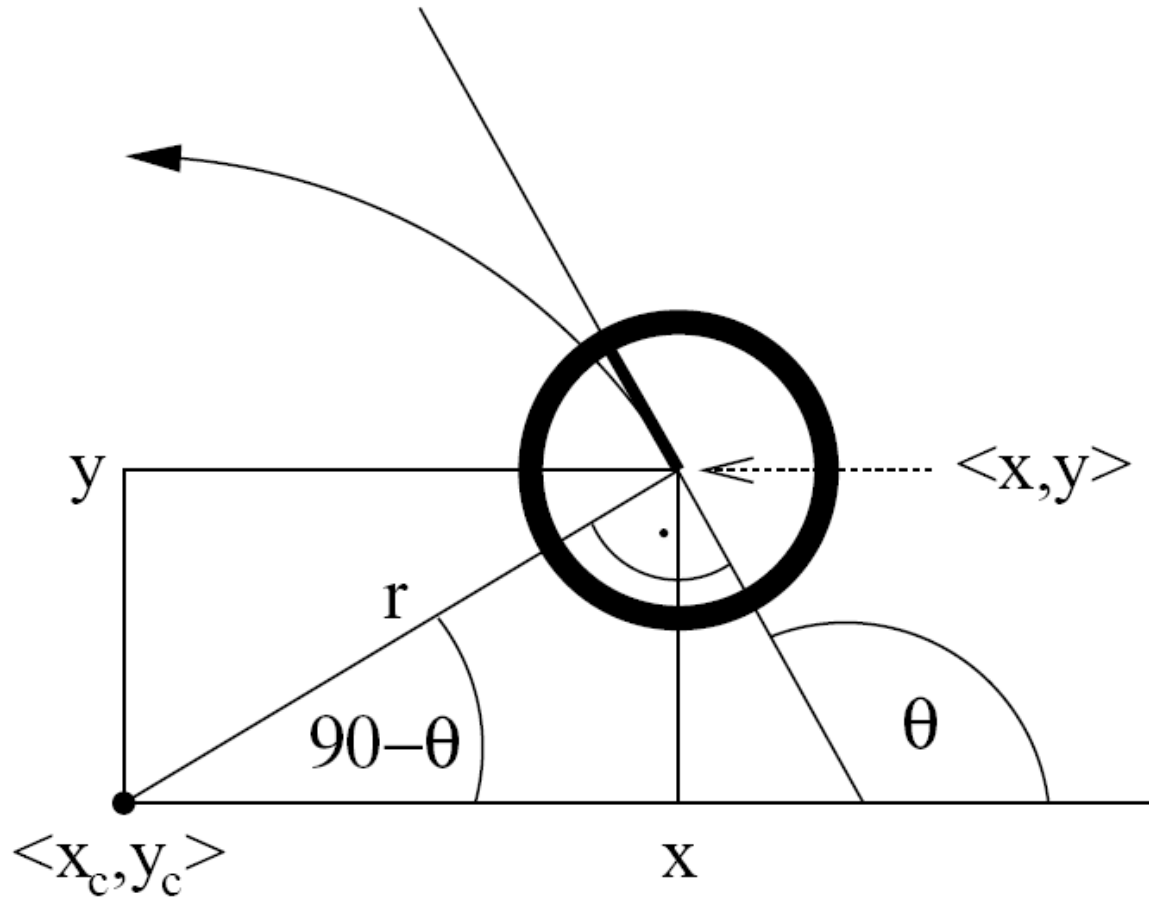
Velocity Motion Model

- ▶ assumes the robot can be controlled through two velocities
 - ▶ translational velocity v
 - ▶ rotational velocity ω
- ▶ our motion command, or control vector, is

$$u_t = \begin{pmatrix} v_t \\ \omega_t \end{pmatrix}$$

- ▶ positive values correspond to forward translation and counterclockwise rotation

Velocity Motion Model



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** $\text{prob}(v - \hat{v}, \alpha_1|v| + \alpha_2|\omega|) \cdot \text{prob}(\omega - \hat{\omega}, \alpha_3|v| + \alpha_4|\omega|)$
 $\cdot \text{prob}(\hat{\gamma}, \alpha_5|v| + \alpha_6|\omega|)$