

# Mobile Robot Navigation Amidst Humans with Intents and Uncertainties: A Time Scaled Collision cone Approach

Akhil Nagariya<sup>1</sup>   Bharath Gopalakrishna<sup>1</sup>   Arun Singh<sup>2</sup>  
Krishnam Gupta<sup>1</sup>   K Madhava Krishna<sup>1</sup>

<sup>1</sup>RRC  
IIIT Hyderabad

<sup>2</sup>Ben-Gurion University, Isreal

CDC 2015

# Outline

Motivation

Human Intention prediction

Proactive collision avoidance in intent space

# Motivation

- ▶ Robots and humans are beginning to occupy the same work spaces
- ▶ Account for human intent in robot's navigation and avoidance Maneuver
- ▶ Uncertain and Haphazard local movements of human

# Outline

Motivation

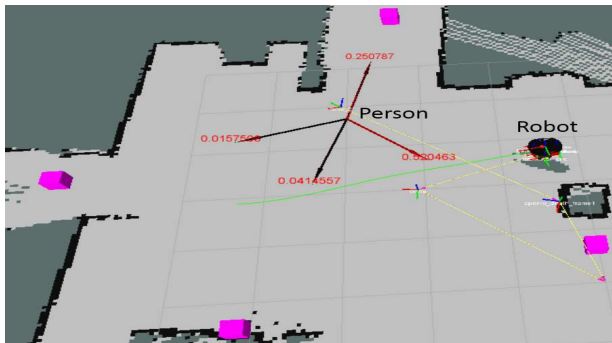
Human Intention prediction

Proactive collision avoidance in intent space

# Human Intention prediction

- ▶ Characterize intents as the final destinations a person might reach
- ▶ Let  $D = \{\mathbf{d}^1, \mathbf{d}^2, \dots, \mathbf{d}^m\}$  be the set of final destinations a person can go to in a given environment
- ▶ compute the probability of each of these intents Using Hidden Markov Model.
- ▶ Characterize local Haphazard movements as a gaussian  $\mathcal{N}(\mu_i(\mathbf{x}^t), \sigma_t)$

# Human Intention prediction

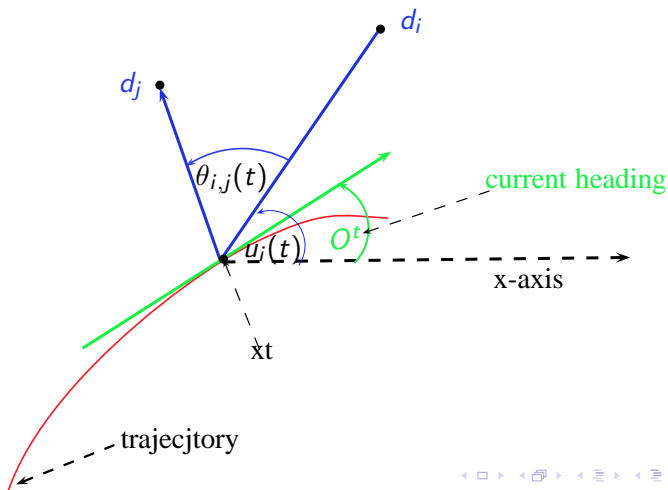


# HMM for Intention prediction

- ▶ Let  $S^t \in D$  represent the intent of a person to reach destination  $S^t$  at time  $t$ .
- ▶  $D$  represents set of states in HMM.
- ▶ Human trajectories are represented as  $X(T) = \{\mathbf{x}^1, \mathbf{x}^2, \dots, \mathbf{x}^T\}$

# HMM for Intention prediction

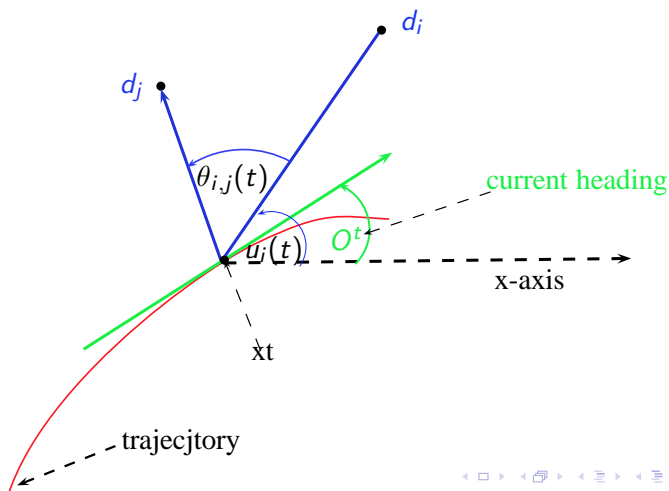
- ▶  $O^t$  is the angle defined by the first derivative of the trajectory at point  $\mathbf{x}^t$
- ▶ Given the current position and orientation we compute the probability of reaching each of the destination  $d^i \in D$





# HMM for Intention prediction

- ▶  $\mu_i(t)$  is the measure relative to the destination  $\mathbf{d}^i$
- ▶  $O^t$  is the global measure of the target orientation
- ▶  $\theta_{ij}(t)$  is the measure between final destinations  $\mathbf{d}^i$  and  $\mathbf{d}^j$  relative to the current position  $\mathbf{x}^t$



# HMM for Intention prediction

- ▶  $b_i(O^t)$  is the probability of observing heading  $O^t$  given that the person is following the intent  $\mathbf{d}^i$  at time  $t$ .

$$b_i(O^t) = p(O^t | S^t = \mathbf{d}^i) = \mathcal{N}(O^t | \mu_i(t), \sigma_o)$$

- ▶  $a_{ij}(t)$  is the probability that the human changes his intent from  $\mathbf{d}^i$  to  $\mathbf{d}^j$  at any discrete instant  $t$

$$a_{ij}(t) = p(S^{t+1} = \mathbf{d}^j | S^t = \mathbf{d}^i) = \eta \mathcal{N}(\theta_{ij}(t) | 0, \sigma_a)$$

# HMM for Intention prediction

- ▶ Let  $O^{1:T} = \{O^1, O^1, \dots, O^T\}$  is the set of measurements obtained till time  $T$ .
- ▶ Our task is to calculate  $p(S^t = \mathbf{d}^i | O^{1:T}, \lambda)$
- ▶ In HMM this term is usually referred to as  $\gamma_t(i)$  To find this we use standard forward and backward algorithms.

# Proactive collision avoidance in intent space

# Summary

- ▶ The **first main message** of your talk in one or two lines.
- ▶ The **second main message** of your talk in one or two lines.
- ▶ Perhaps a **third message**, but not more than that.
- ▶ Outlook
  - ▶ Something you haven't solved.
  - ▶ Something else you haven't solved.

# For Further Reading I



A. Author.

*Handbook of Everything.*

Some Press, 1990.



S. Someone.

On this and that.

*Journal of This and That*, 2(1):50–100, 2000.