

# CS:4980

# Foundations of Embedded Systems

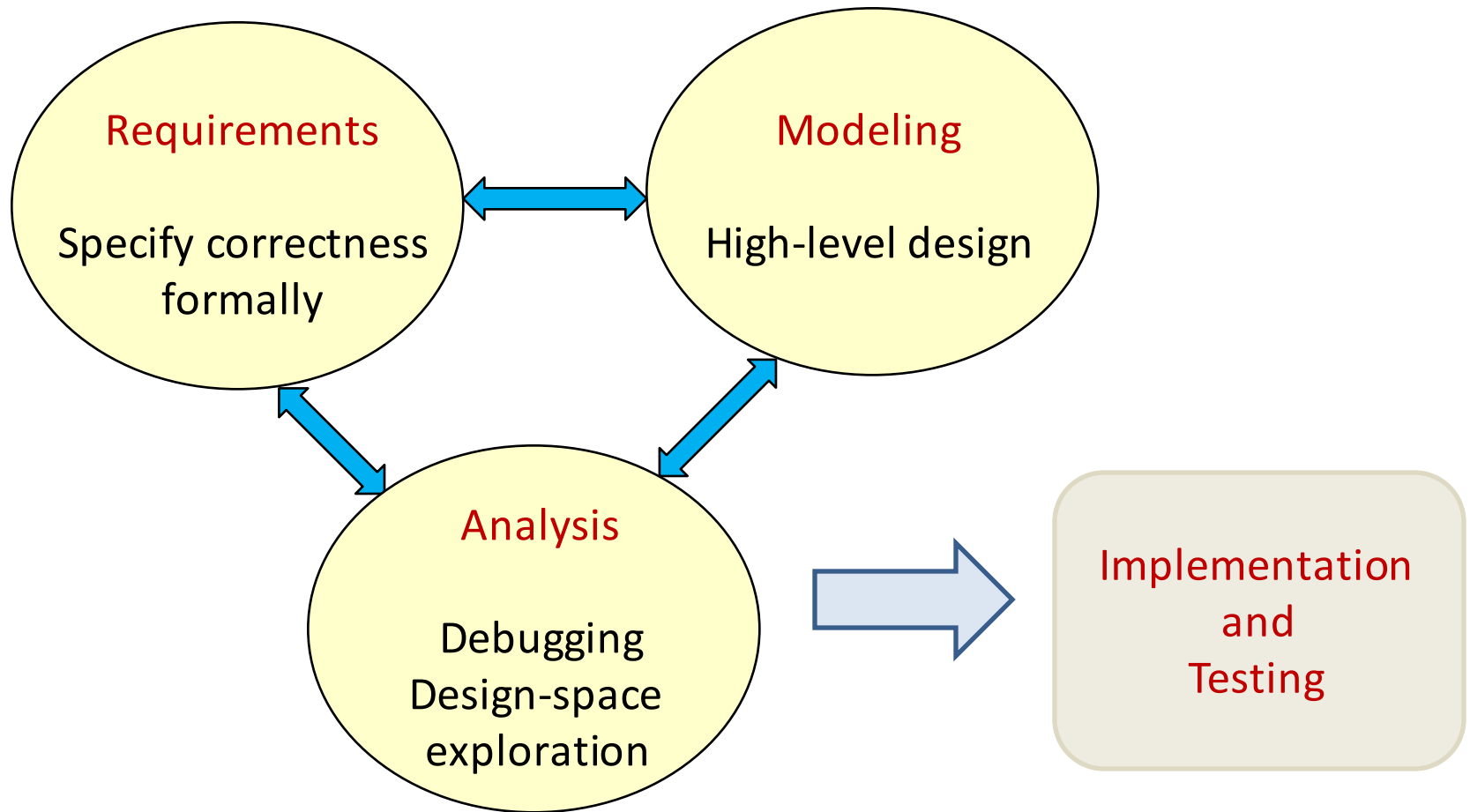
## Hybrid Systems

## Part II

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# Model-Based Design and Analysis



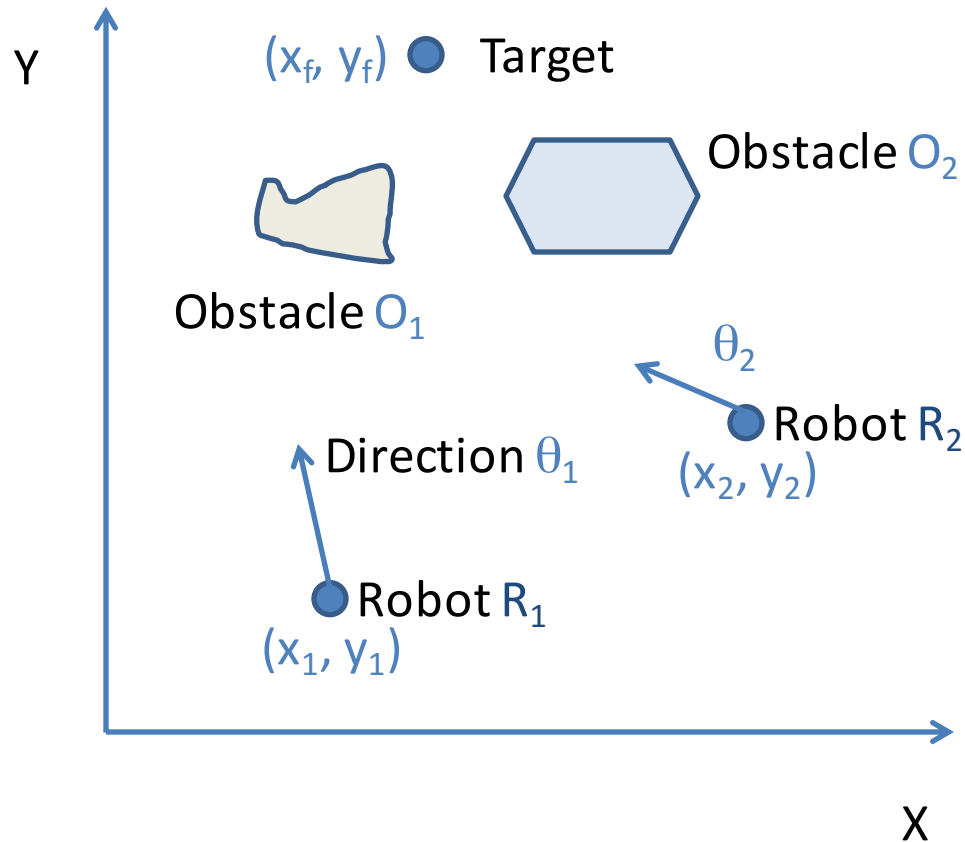
# Multi-Robot Coordination

- ❑ Autonomous mobile robots in a room
- ❑ Goal of each robot:
  - Reach a target at a known location
  - Avoid obstacles (positions of obstacles not known in advance)
  - Minimize distance travelled
- ❑ Cameras and vision processing algorithms allow each robot to estimate obstacle positions
  - Estimates are only approximate, and depend on relative position of obstacles with respect to a robot's position
  - How often should robot update these estimates ?

# Multi-Robot Coordination

- ❑ Each robot can communicate with others using wireless links
  - How often and what information?
  - How does communication help?
- ❑ High-level motion control (path planning)
  - Decide on speed and direction

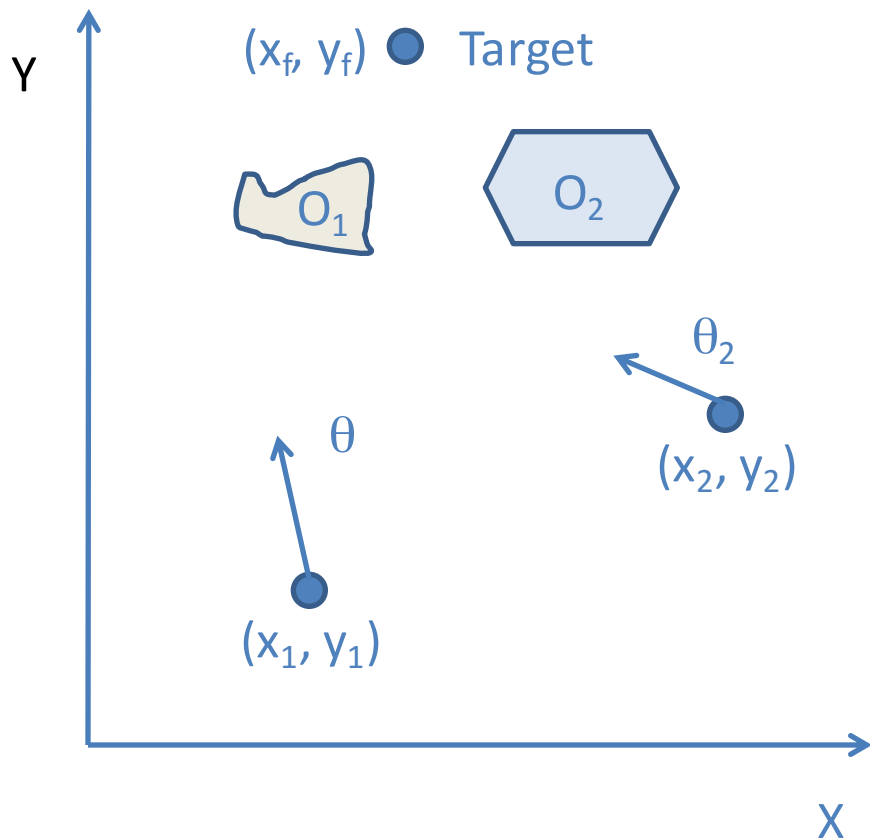
# Path Planning with Obstacle Avoidance



Assumptions:

- Two dimensional world
- Point robots
- Fixed speed  $v$

# Path Planning with Obstacle Avoidance



State variables:  $(x_1, y_1), (x_2, y_2)$

Initialization:

$$(x_1, y_1) := (x_{10}, y_{10})$$

$$(x_2, y_2) := (x_{20}, y_{20})$$

Dynamics:

$$dx_1 = v \cos \theta \quad dx_2 = v \cos \theta_2$$

$$dy_1 = v \sin \theta \quad dy_2 = v \sin \theta_2$$

Safety requirement:

$$(x_1, y_1), (x_2, y_2) \notin O_1 \cup O_2$$

Liveness requirement:

Eventually  $(x_1, y_1) = (x_f, y_f)$  and

Eventually  $(x_2, y_2) = (x_f, y_f)$

Performance: Reduce distance travelled!

# Abstractions

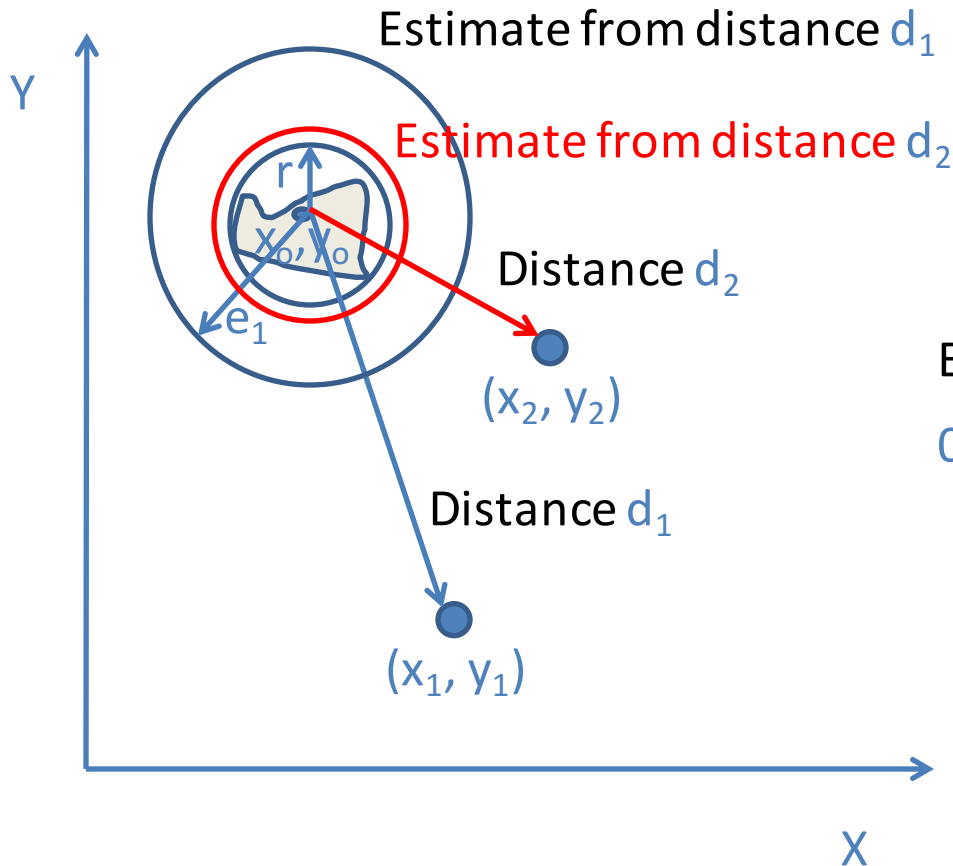
- ❑ For modeling and analysis for motion planning, we need to **simplify** obstacle shapes and complexity of image processing algorithms
  - **Simplicity and abstraction**: key to modeling
- ❑ Assume each robot is a **point**
  - Can be described by coordinates of point
- ❑ Assume each obstacle/estimate is a **circle**
  - Can be described by coordinates of center and radius
  - Assumption: real obstacle is always contained in estimated circle
  - Alternative: ellipses (more accurate)

# Modeling Obstacles

- ❑ Consider an obstacle with center  $(x_o, y_o)$  and radius  $r$ 
  - Radius of smallest circle that envelopes the actual obstacle
- ❑ Estimate of the obstacle as computed by a robot using image processing algorithms of a robot
  - A circle with center  $(x_o, y_o)$  and radius  $e > r$
  - The closer the robot to the obstacle, the better the estimate
  - Estimate  $e$  decreases with distance of robot from obstacle, and converges to  $r$



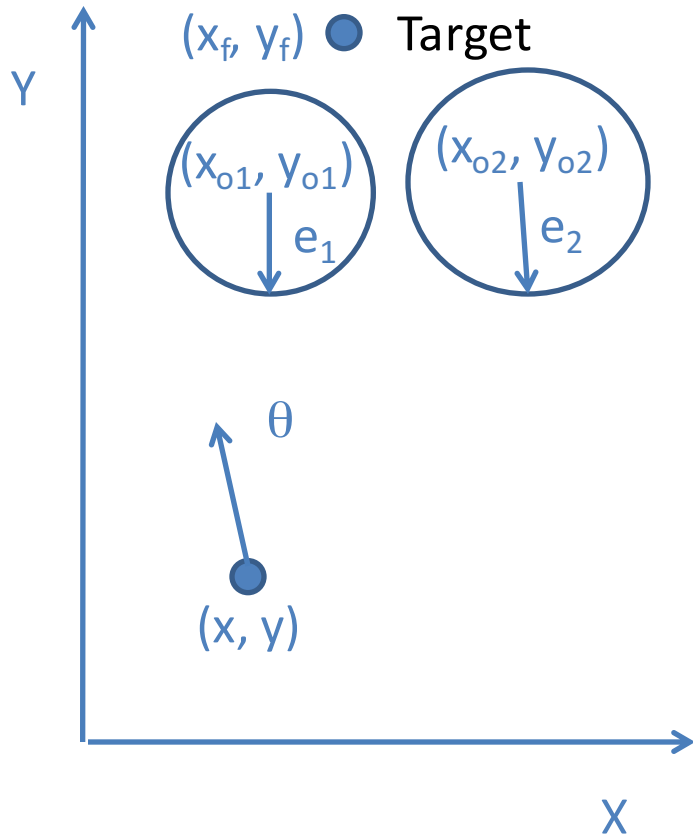
# Obstacle Estimation



Estimated radius:  $e_1 = r + a (d_1 - r)$

$0 < a < 1$  is a constant

# Rule for Obstacle Estimation



Robot  $R_1$  maintains radii  $e_1$  and  $e_2$  that are estimates of the obstacles

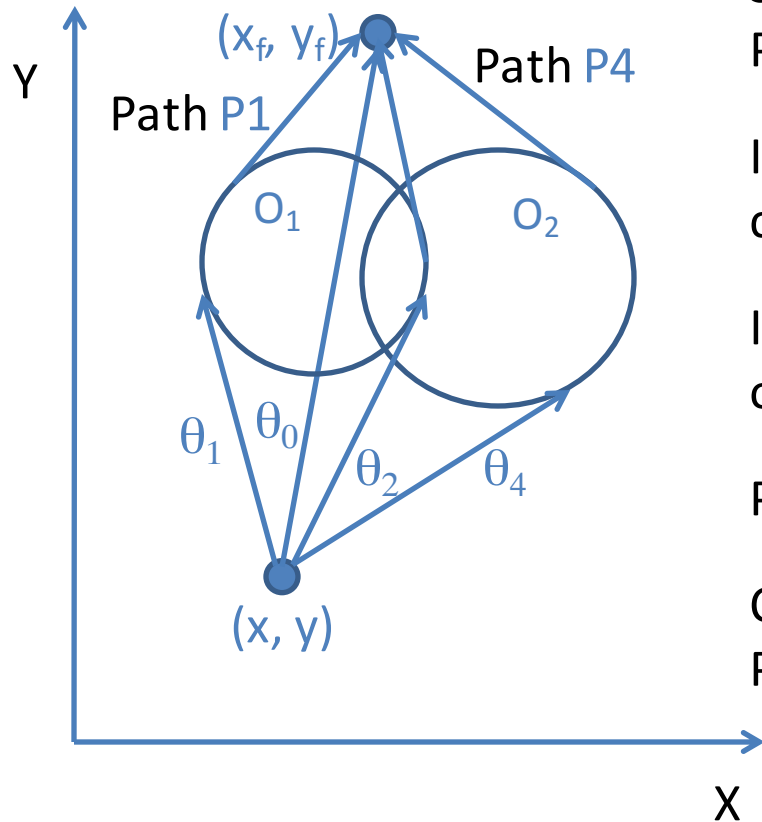
Obstacle estimation is computationally expensive

Every  $t_e$  seconds, robot executes discrete update:

$$e_1 := \min(e_1, r_1 + a(\text{dist}((x, y), (x_{o1}, y_{o1})) - r_1) ;$$
$$e_2 := \min(e_2, r_2 + a(\text{dist}((x, y), (x_{o2}, y_{o2})) - r_2)$$

Computation for robot  $R_2$  is symmetric

# Path Planning



Shortest path: straight line to target

Preferred direction:  $\theta_0$

If estimate of obstacle  $O_1$  intersects straight path, calculate two paths that are tangents to obstacle

If estimate of obstacle  $O_2$  intersects straight path, or obstacle  $O_1$ , calculate tangent paths

Plausible paths: **P1** and **P4**

Calculate which one is shorter:

Planning algorithm returns either  $\theta_1$  or  $\theta_4$

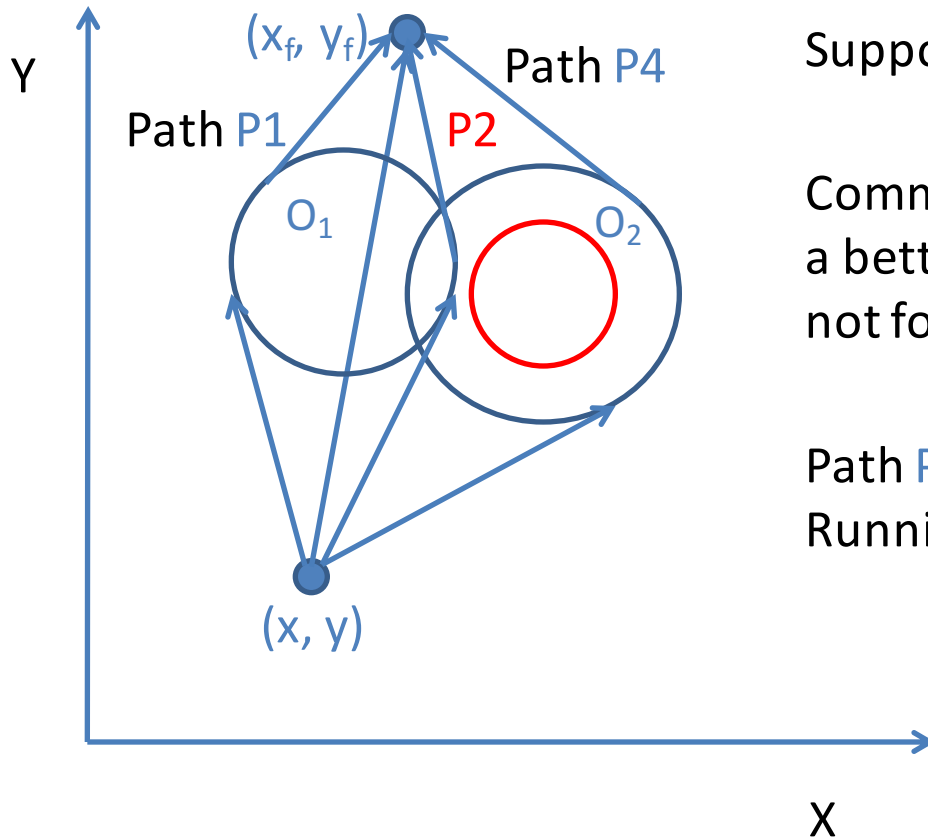
# Path Planning

- ❑ Function plan with inputs:
  - current position of robot  $R_i$
  - target position
  - obstacle  $O_1$  position (center and radius estimate)
  - obstacle  $O_2$  position (center and radius estimate)
  
- ❑ Output: Direction for motion
  - Best possible path to target while avoiding obstacles and assuming estimates are correct
  
- ❑ Function plan written in C code (can be embedded in model)
  
- ❑ Does it help to rerun planning algorithm again as robot moves?
  - Yes! Estimates may improve, suggesting shorter paths
  - Invoke planning algorithm every  $t_p$  seconds

# Communication

- ❑ Each robot has its own estimate of each obstacle
- ❑ Robot  $R_2$ 's estimates may be better than  $R_1$ 's own estimates
- ❑ Strategy: Every  $t_c$  seconds, send your own estimates to the other robot, and receive estimates from it
- ❑ If your own estimates are  $e_{i1}$  and  $e_{i2}$ , and you receive estimates  $e_{j1}$  and  $e_{j2}$ , set
$$e_{i1} := \min(e_{i1}, e_{j1})$$
$$e_{i2} := \min(e_{i2}, e_{j2})$$

# Effect of Coordination



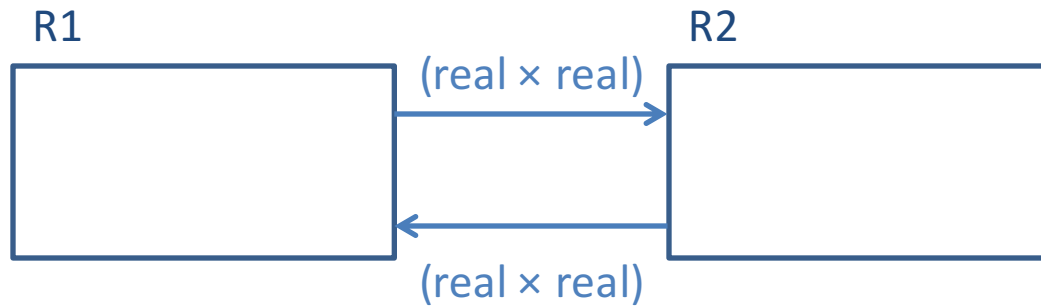
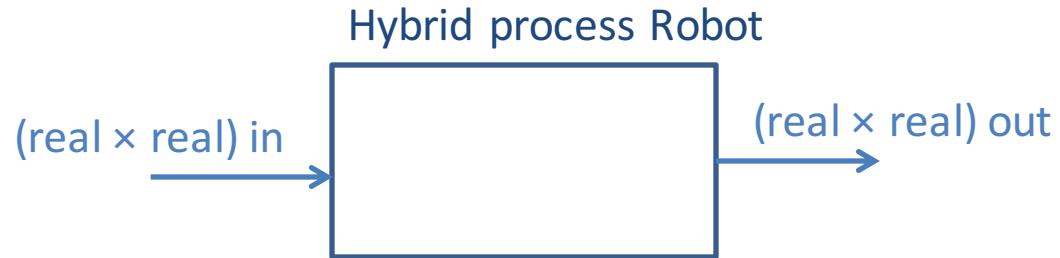
Suppose Path **P1** was preferred

Communication with other robot gives a better estimate of obstacle  $O_2$ , but not for obstacle  $O_1$

Path **P2** is now viable.

Running planner again could choose path **P2**

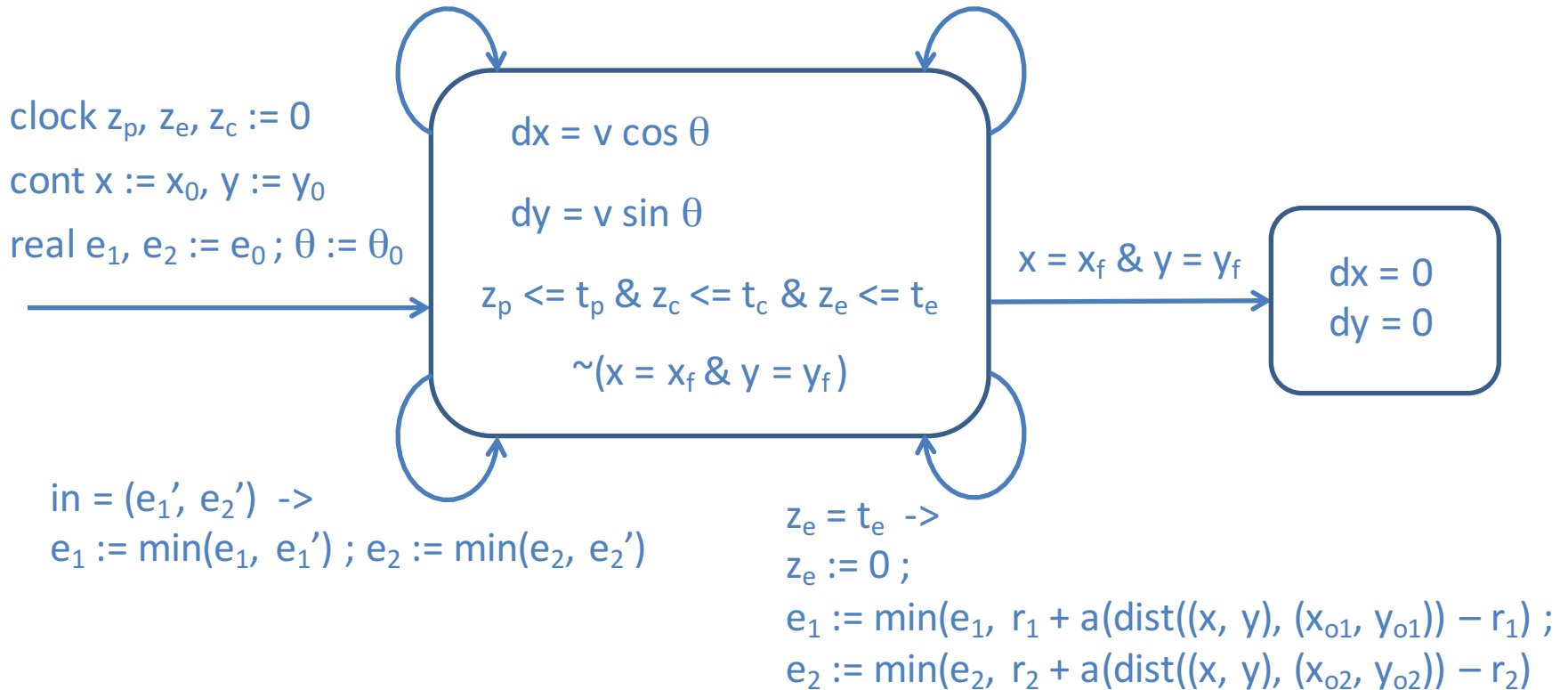
# System of Robots



# Robot Model

$z_c = t_c \rightarrow \text{out} := (e_1, e_2) ; z_c := 0$

$z_p = t_p \rightarrow \theta := \text{plan}(x, y, e_1, e_2) ; z_p := 0$

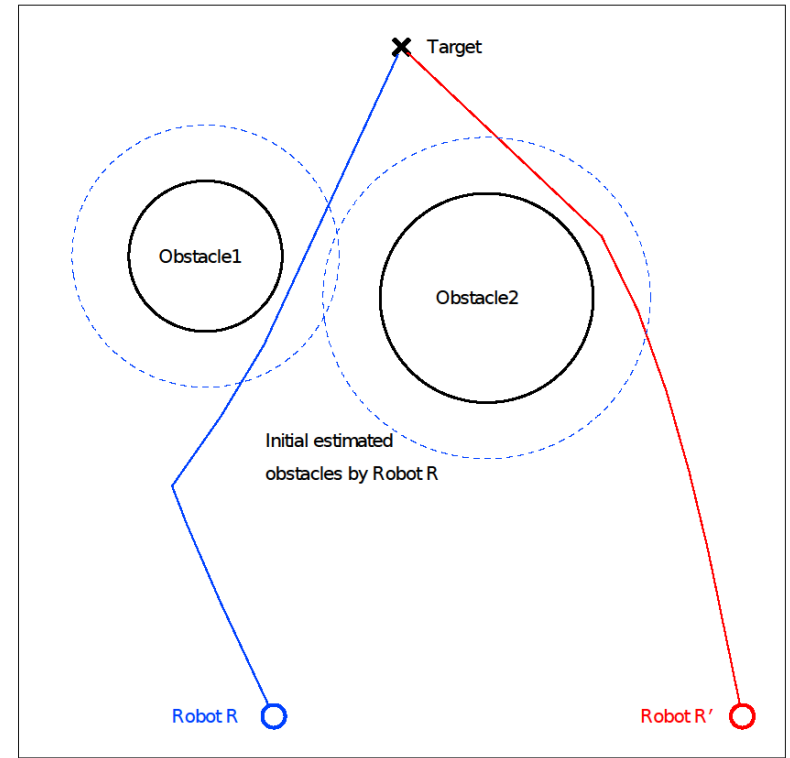
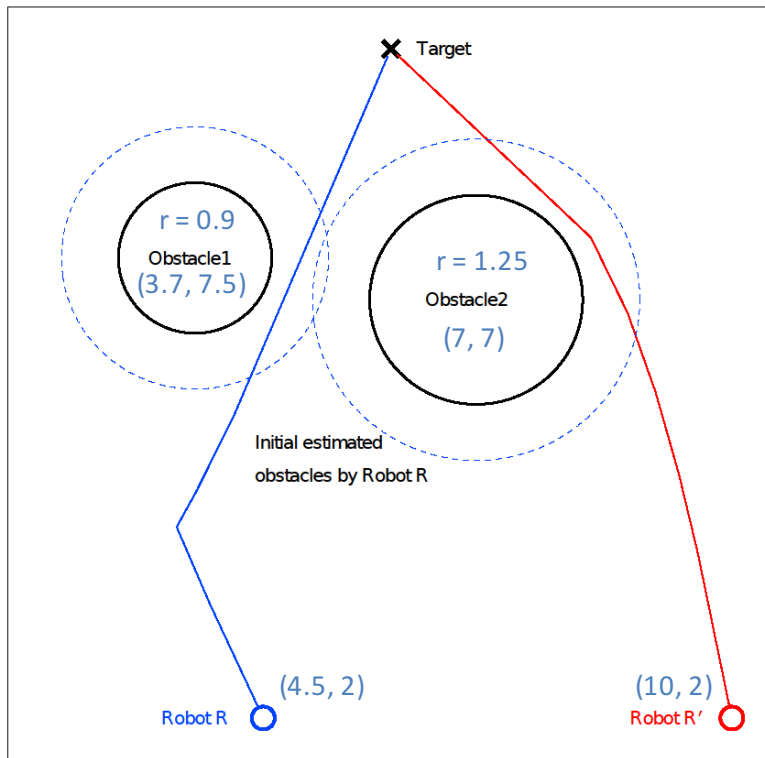




# Analysis

- ❑ Key system parameters
  - How often should a robot communicate?
  - How often should a robot execute planning algorithm
  - How often should a robot execute image processing algorithm to update obstacle estimates?
  
- ❑ Design-space exploration: Choose values of  $t_c$ ,  $t_p$ ,  $t_e$ 
  - Reduce distance travelled, but also account for costs of communication/computation
  
- ❑ Symbolic analysis beyond the scope of current tools, so need to run multiple simulations

# Illustrative Execution

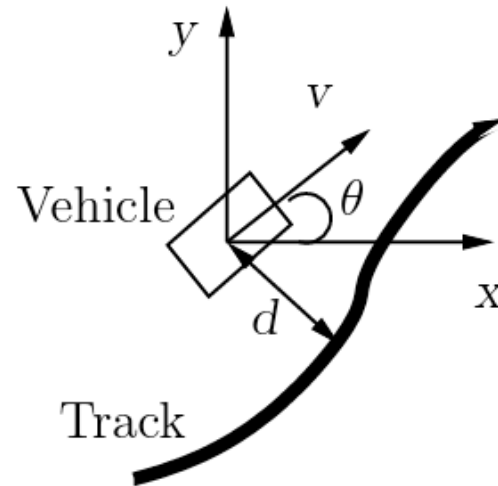


- Speed  $v$  :  $0.5 \text{ u/s}$
- Planning rate  $t_p$  :  $2 \text{ s}$
- Obstacle estimation rate  $t_p$  :  $2 \text{ s}$
- Communication rate  $t_c$  :  $4 \text{ s}$
- Distance travelled by R' :  $9.15 \text{ u}$
- Distance travelled by R :  $8.65 \text{ u}$

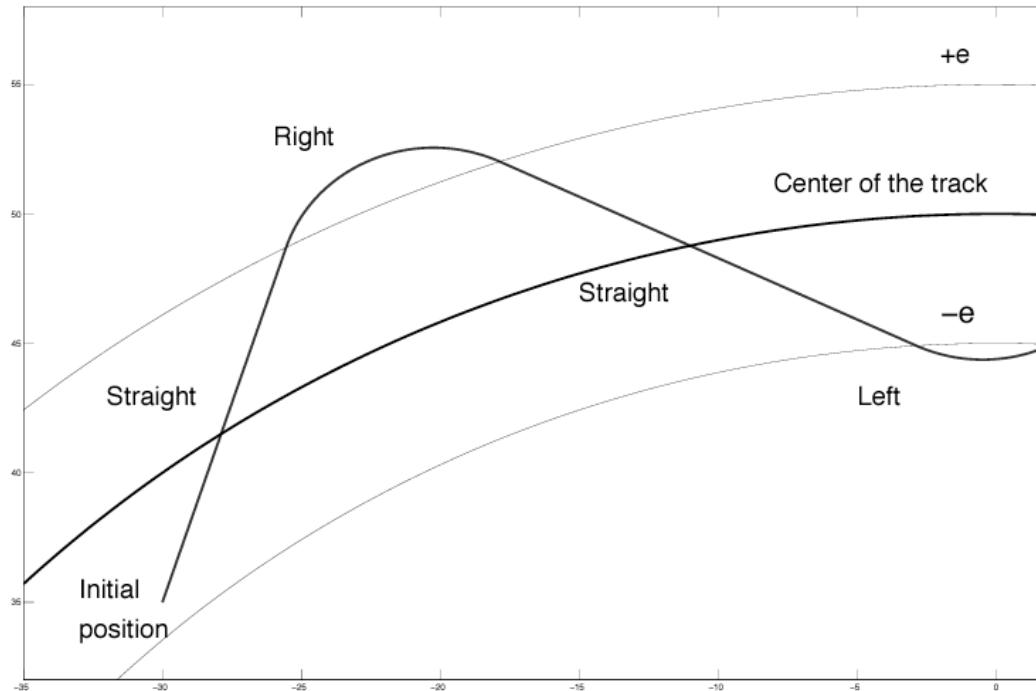
- Speed  $v$  :  $0.5 \text{ u/s}$
- Planning rate  $t_p$  :  $2 \text{ s}$
- Obstacle estimation rate  $t_p$  :  $2 \text{ s}$
- Communication rate  $t_c$  :  $\gg 4 \text{ s}$
- Distance travelled by R' :  $9.15 \text{ u}$
- Distance travelled by R :  $8.81 \text{ u}$

# Automated Guided Vehicle

- ❑ Autonomous vehicle on a flat surface, following a visual track
- ❑ Goal of each robot:
  - Move along a track (i.e., center line of a road)
  - Follow track as close as possible
- ❑ Cameras and vision processing algorithms allow vehicle to sense track and measure (signed) distance  $d$  from center of the track
- ❑ Two degrees of freedom: move forward and rotate
- ❑ Two velocities: (regular) velocity  $(v, \theta)$  and angular velocity  $\omega$



# Automated Guided Vehicle Controller



**Inputs:**  $\{\text{start}, \text{stop}\}$  command  $c$ , distance  $d$  from center of track

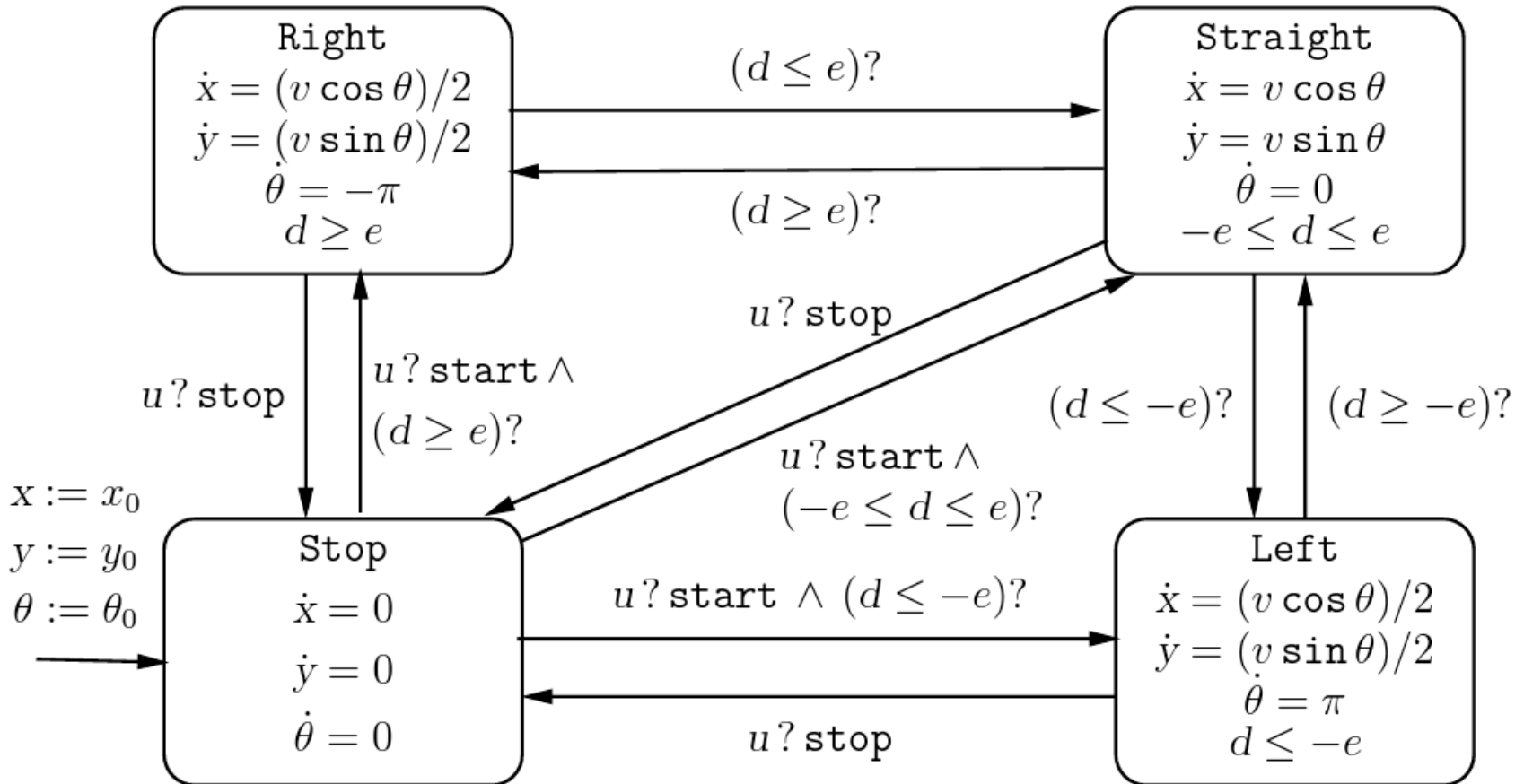
**Outputs:** speed  $v$ , angular speed  $\omega$

**State:** coordinates  $x, y$ ; angle  $\theta$

**Modes:** Stop, Straight, Left, Right

**Simplifications:**  $v \in \{v_c/2, v_c\}$  and  $\omega \in \{-\pi, 0, \pi\}$

# Automated Guided Vehicle Controller



# Credits

Notes based on Chapter 9 of

## **Principles of Cyber-Physical Systems**

by Rajeev Alur

MIT Press, 2015