



# Symbolic Planning for Dynamic Robots

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# TECHLAV Project

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- Testing, Evaluation and Control of Heterogeneous Large-scale Autonomous systems of Vehicles (TECHLAV)▫
- Thrust 1: Modeling, Analysis and Control of Large-scale Autonomous Vehicles (MACLAV)▫
- Task 1-5: Hierarchical Hybrid Cooperative Control of LSASV▫



# TECHLAV Project

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## Objective

Developing capable and scalable models for autonomous collaborative, robust and distributed decision-making, group coordination, planning, and tasking through effective interaction with human operators

## Impact

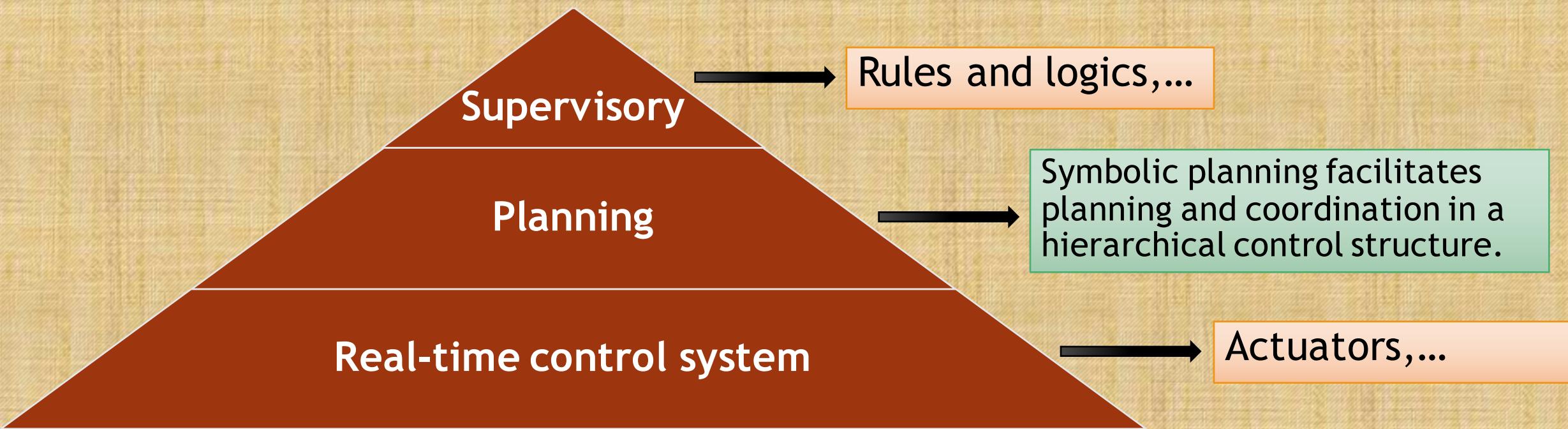
Reaching higher levels of autonomy and teaming of multi-agent systems by using agents which are able to accomplish assigned missions autonomously and have the capability of autonomous collaboration with other teammates and human operators.

# Relation to TECHLAV

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Task 1-5: Hierarchical Hybrid Cooperative Control of LSASV

## Hierarchical Control



# Outline

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- 1 • Robot Motion Planning and Control
- 2 • Task Specification
- 3 • Temporal Logic
- 4 • Symbolic Planning
- 5 • Future Work

# Mobile Robots

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## Components

- mechanical elements (wheels and gears)
- electromechanical devices (motors, clutches and brakes)
- digital circuits (processors and smart sensors)
- software programs (embedded controllers)

## Constraints

- mechanical constraints (e.g., a car-like robot cannot move sideways)
- limited energy resources, and computation, sensing, and communication capabilities.

## Complexities

- Environment is cluttered with possibly moving and shape changing obstacles.
- Their objectives can change over time, such as in the case of appearing and disappearing targets.

# New Task Requirements for Robots

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- Converging to a desired operating point while always staying within a safe set
- Executing sequenced tasks
- Reaching certain areas and visiting certain areas infinitely often
- Avoiding obstacles

# Robot Motion Planning and Control

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## Problem

- automatic construction of robot control strategies from task specifications given in high-level, human-like language

## Challenge

- the development of frameworks for control design which is:
  - computationally efficient
  - allowing for systematic, provably correct, control design
  - accommodating both the robot constraints and the complexity of the environment
  - allowing for expressive human-like task specifications

# Task Planning

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## Mission requirements:

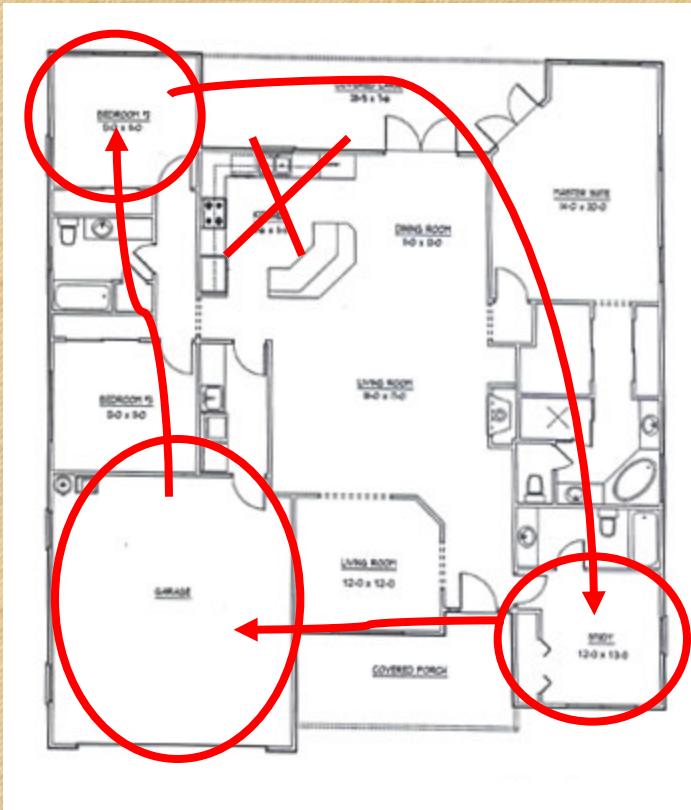
- converging to a desired operating point while always staying within a safe set
- executing sequenced tasks
- reaching certain areas and visiting certain areas infinitely often
- avoiding obstacles

How to formally describe the task?

Temporal logic

# Why Temporal Logic?

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- A formal high level language to describe a complex mission
- A wider class of properties than safety and stability
- Having well defined syntax and semantics, which can be easily used to specify complex behavior

# Introduction to Temporal Logic

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**Temporal logic:**

Any system of rules and symbolism for representing, and reasoning about, propositions qualified in terms of time.

**Classical logic:**

- “It is Monday”

**Temporal logic**

- “I am *always* hungry”
- “I will *eventually* be hungry”
- “I will be hungry *until* I eat something”

# Temporal Operators

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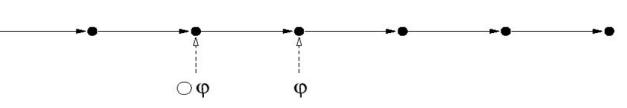
## Operators

## Definition

## Diagram

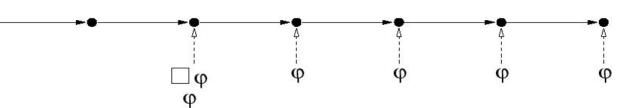
$O\varphi$

$\varphi$  is true in the next moment of time



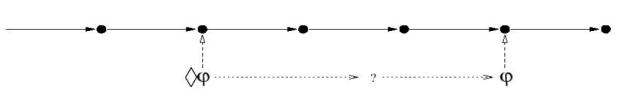
$\Box\varphi$

$\varphi$  is true in all future moments



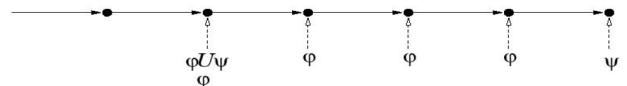
$\Diamond\varphi$

$\varphi$  is true in some future moment



$\varphi u \psi$

$\varphi$  is true until  $\psi$  is true



# Temporal Syntax

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$$\varphi ::= p \mid \neg\varphi \mid \varphi \wedge \varphi \mid \varphi \vee \varphi \mid \varphi \rightarrow \varphi$$

$p$	an arbitrary atomic proposition
$\neg$	Negation (complement)
$\wedge$	conjunction
$\vee$	disjunction
$\rightarrow$	implication

**Example:**  
 $(p \wedge \neg q) \rightarrow r$

# Temporal Semantics

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- A tool for specification, formal analysis, and verification of the executions of computer programs and systems

$\mathcal{M} = \langle \mathbb{N}, I \rangle$	a discrete, linear model of time
$I: \mathbb{N} \rightarrow 2^\Sigma$	maps each natural number to a set of propositions
$\models: (\mathcal{M} \times \mathbb{N} \times FORM) \rightarrow \{\text{true}, \text{false}\}$	Satisfaction relation

$\langle \mathcal{M}, i \rangle \models p \text{ iff } p \in I(i)$  (for  $p \in \Sigma$ )

$$\begin{aligned}
 \langle \mathcal{M}, i \rangle \models \neg\varphi &\quad \text{iff} \quad \langle \mathcal{M}, i \rangle \not\models \varphi \\
 \langle \mathcal{M}, i \rangle \models \varphi \wedge \psi &\quad \text{iff} \quad \langle \mathcal{M}, i \rangle \models \varphi \text{ and } \langle \mathcal{M}, i \rangle \models \psi \\
 \langle \mathcal{M}, i \rangle \models \varphi \vee \psi &\quad \text{iff} \quad \langle \mathcal{M}, i \rangle \models \varphi \text{ or } \langle \mathcal{M}, i \rangle \models \psi \\
 \langle \mathcal{M}, i \rangle \models \varphi \Rightarrow \psi &\quad \text{iff} \quad \text{if } \langle \mathcal{M}, i \rangle \models \varphi \text{ then } \langle \mathcal{M}, i \rangle \models \psi \\
 \mathcal{M}, i \models \top & \\
 \mathcal{M}, i \not\models \perp &
 \end{aligned}$$

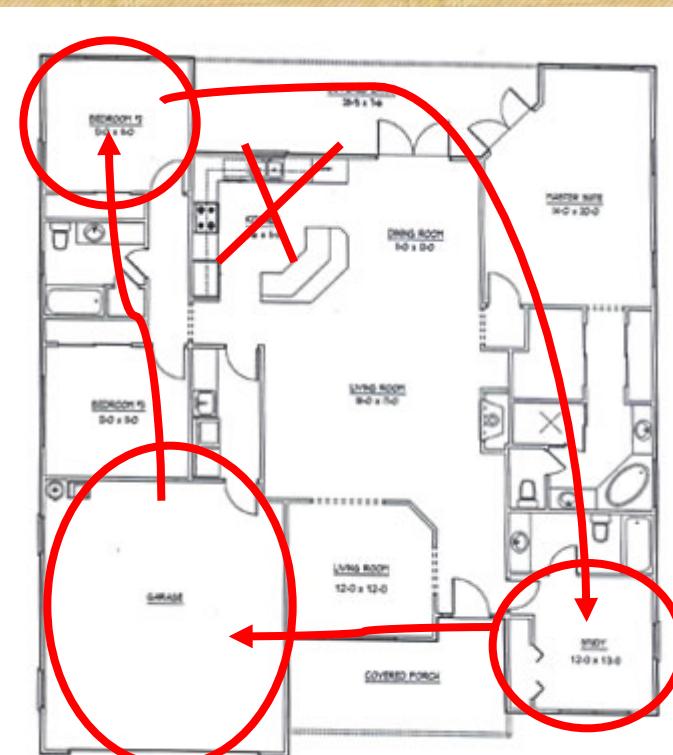
# LTL for Robot Task Specifications

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- **Coverage:** eventually visit all regions
  - **Sequencing:** visit P2 before you go to P3
  - **Avoidance:** until you go to P2 avoid P1 and P3

## Examples:

1. “visit rooms 1,2,3 while avoiding corridor 1!”  
 $\varphi = \square\neg(\text{corridor1}) \wedge \square(\text{room1}) \wedge \square(\text{room2}) \wedge \square(\text{room3})$
  2. “if the light is on, visit rooms 1 and 2 infinitely often!”  
 $\varphi = \square(\text{lightOn}) \rightarrow (\square\lozenge(\text{room1}) \wedge \square\lozenge(\text{room2}))$
  3. “if you are in room 3 and Mike is there, beep!”  
 $\varphi = \square((\text{room3}) \wedge (\text{SeeMike}) \rightarrow (\text{beep}))$



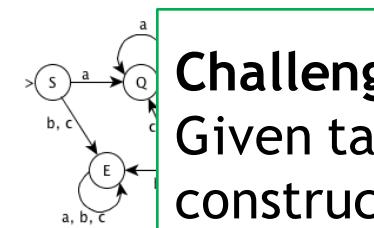
# Symbolic Planning

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## Task

$$\varphi = \square( (\text{room3}) \wedge (\text{SeeMike}) \rightarrow (\text{beep}))$$

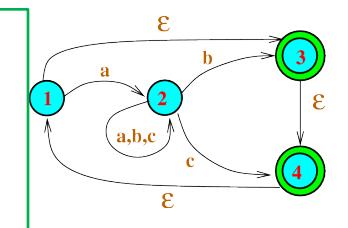
Hybrid Controller



### Challenge:

Given task  $\varphi$  and a dynamical model of the robot  
construct controllers such that:  
 $\text{Robot} \parallel \text{Controller} \models \varphi$

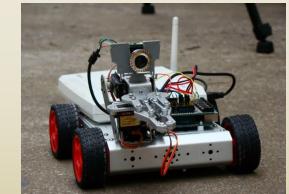
## Abstraction



## Robot

## Interface

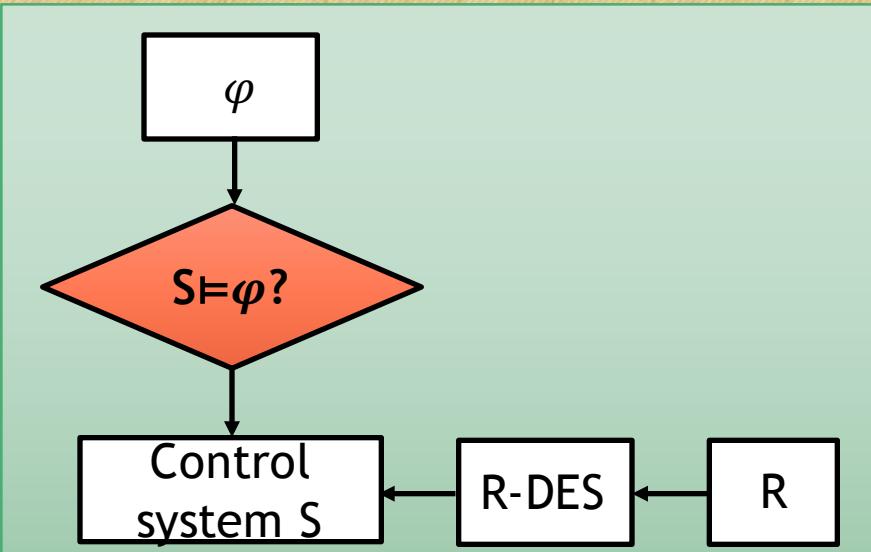
$$\begin{aligned}\dot{x}(t) &= u(t) \\ x(t) &\in P \subset \mathbb{R}^2 \\ u(t) &\in U \subset \mathbb{R}^2\end{aligned}$$



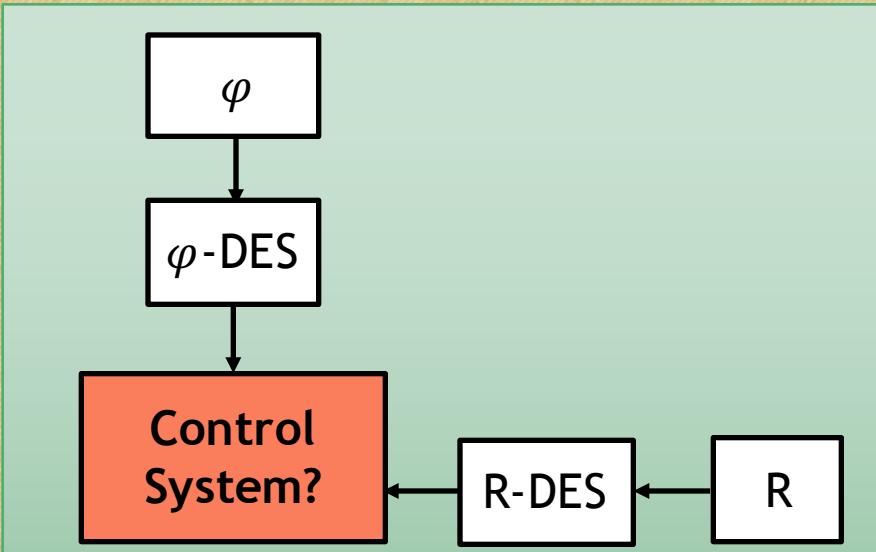
# Control Synthesis

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## Verification

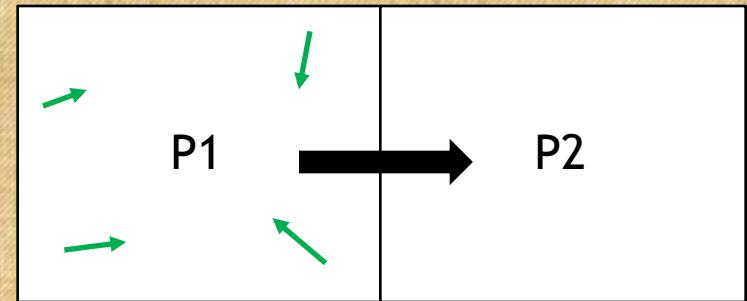


## Planning



# Applying the DES Supervisor to the Robot

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## Challenge:

- DES supervisor tells the robot to go from P1 to P2.
- The planner has no idea about the status of robot.

## Question:

- How to map the high-level task into the continuous controller?

# Bisimulation Equivalency

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- We have to couple equivalency to specification.

Language  
equivalency



- Preserves LTL properties
- Continuous system and DES generate the same trajectories

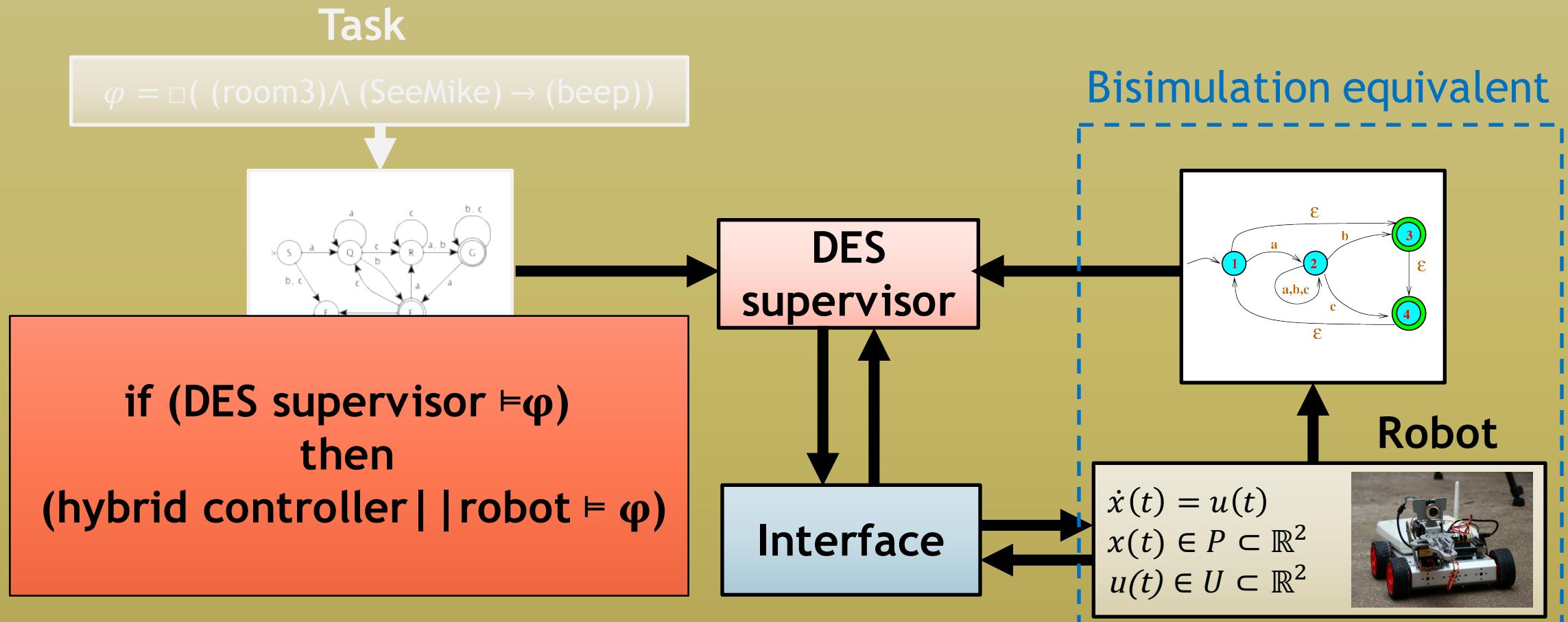
Bisimulation  
equivalency



- Preserves LTL properties
- Continuous system and DES are equivalent in all states (stronger)

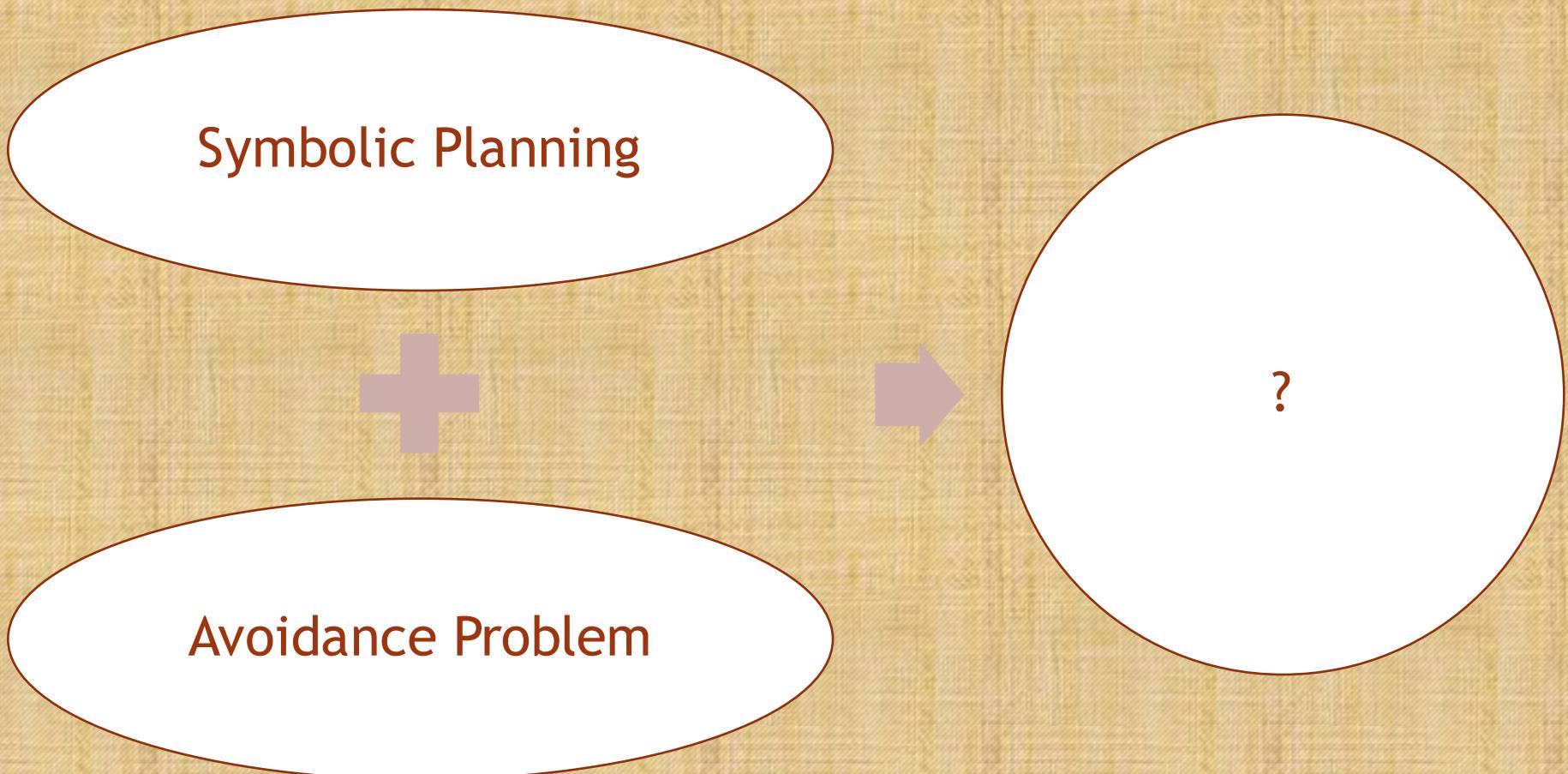
# Symbolic Planning

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# Future Work

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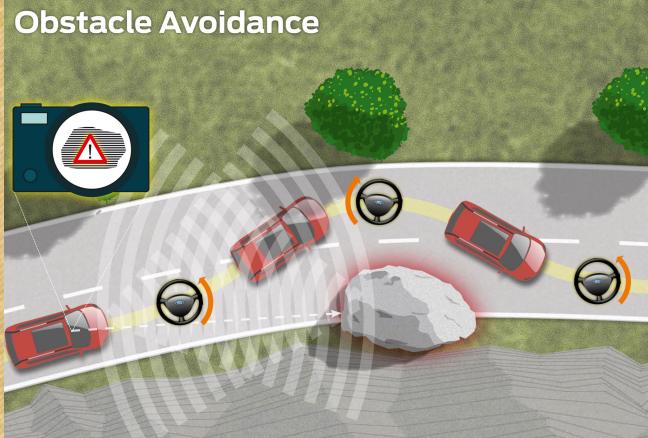


# Future Work

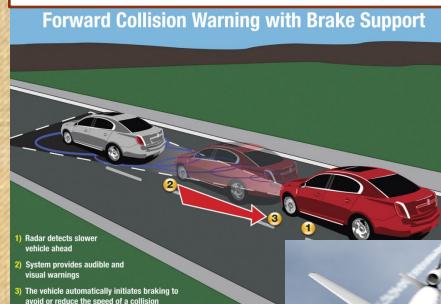
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## Avoidance problem

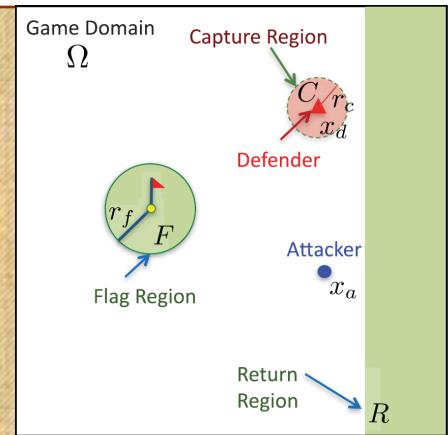
### Obstacle avoidance



### Collision avoidance



### Reach-avoidance



# Future Work

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## Reach-avoidance

### Graph theory

Partitions the graph into regions that are defended and cleared by a team of pursuers to solve simplified path-planning

### Discrete games

Visibility-based pursuit-evasion games: a group of searchers attempt to bring an evader into their field of view, through cooperative coverage of regions

### Path planning with obstacle

Use the possible future positions of the obstacles as static obstacles, allowing using static planning methods(Certain applications with simple configurations of moving adversarial obstacles,).

### model-predictive control

Predicts opponent actions so that optimization can be performed for the controlled agents with respect to the assumed opponent behavior

### Differential game

Winning strategies for the opposing agents can be viewed as the solution to a zero-sum differential game. Simple tasks.

- 1 • Formal description of tasks
- 2 • Efficient computation
- 3 • Combining reaching problem with avoiding problem
- 4 • Targeting more complex tasks (sequencing and ...)

# References

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Thank you 😊

Question?