

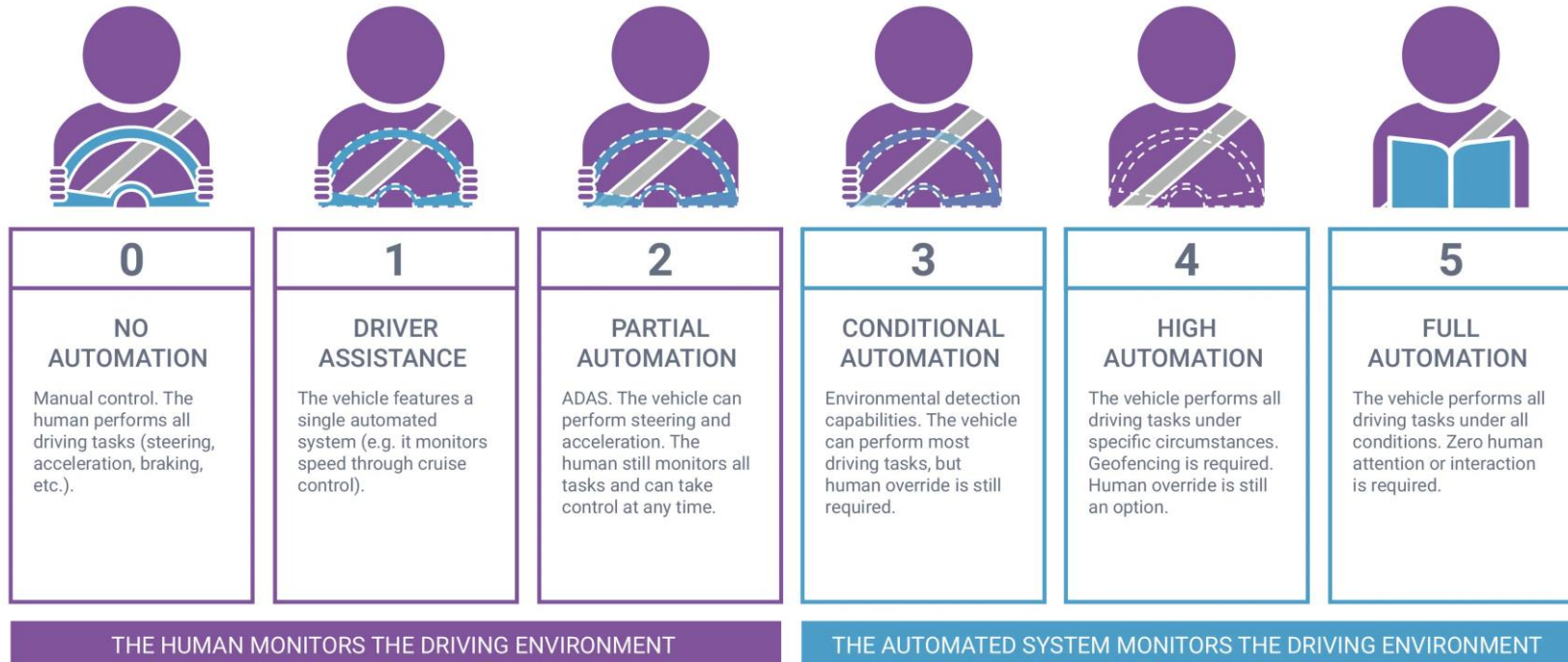


MODULE 4: HUMAN FACTORS IN AUTONOMOUS DRIVING

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SAE International: Levels of Automation





SAE International: Levels of Automation



1 Driver Assistance	2 Partial Automation	3 Conditional Automation	4 High Automation	5 Full Automation
Example <ul style="list-style-type: none">City Emergency Braking	Example <ul style="list-style-type: none">Adaptive Cruise Control	Example <ul style="list-style-type: none">Highway Pilot	Example <ul style="list-style-type: none">City & Highway Pilot	Example <ul style="list-style-type: none">Robocar
Definition <ul style="list-style-type: none">Steering OR acceleration / decelerationHuman supervision	Definition <ul style="list-style-type: none">Steering AND acceleration / decelerationHuman supervision	Definition <ul style="list-style-type: none">All driving functionsHuman intervention may be needed	Definition <ul style="list-style-type: none">All driving functionsHuman can but don't has to intervene	Definition <ul style="list-style-type: none">All driving functionsNo human intervention possible

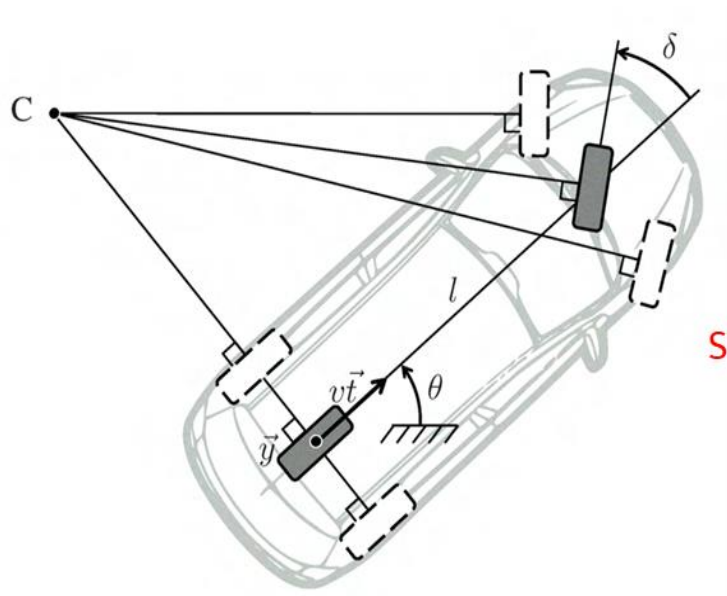


LEVEL 1: DRIVER ASSISTANCE

A SINGLE AUTOMATED SYSTEM

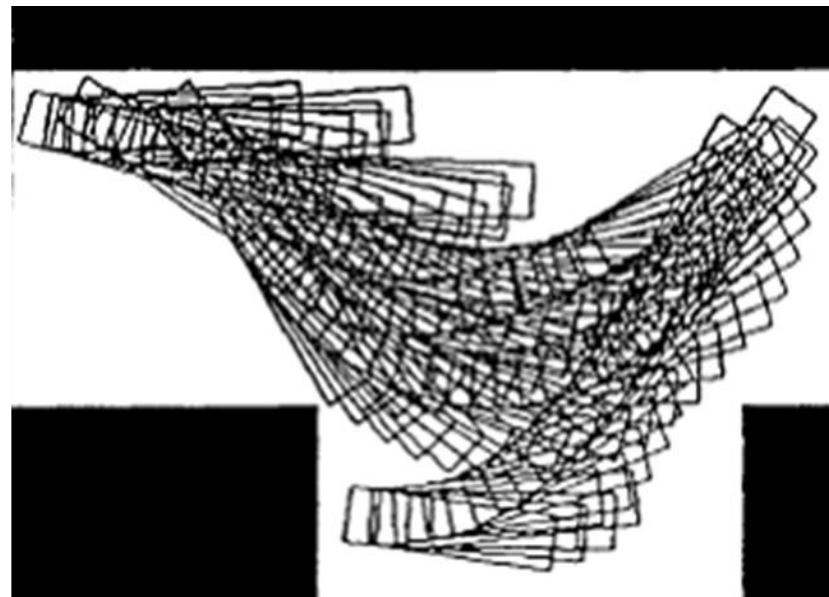


Nonholonomic car model



Steering

Acceleration





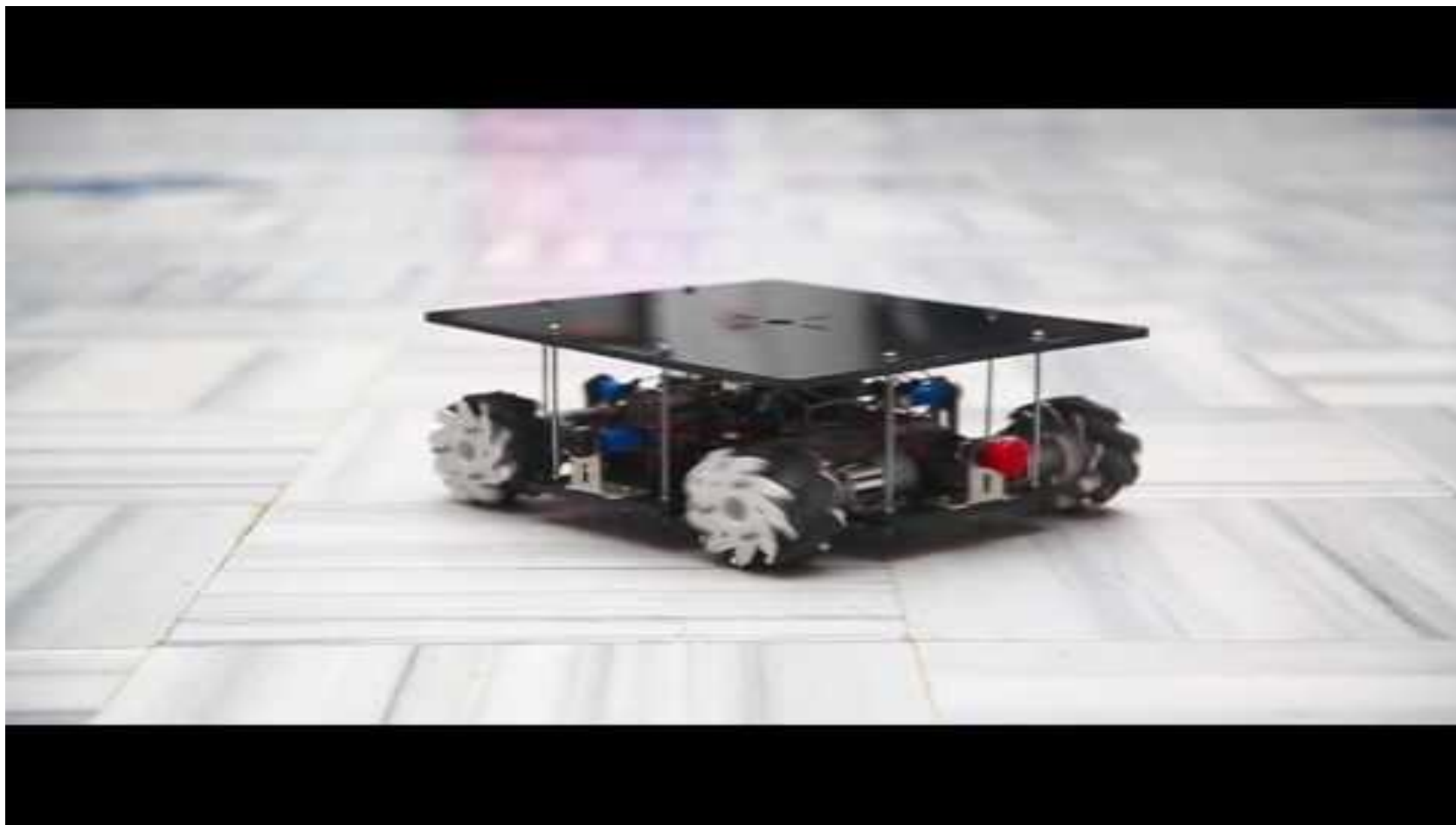
Holonomic robot model

Holonomic system where a robot can move in any direction in the configuration space





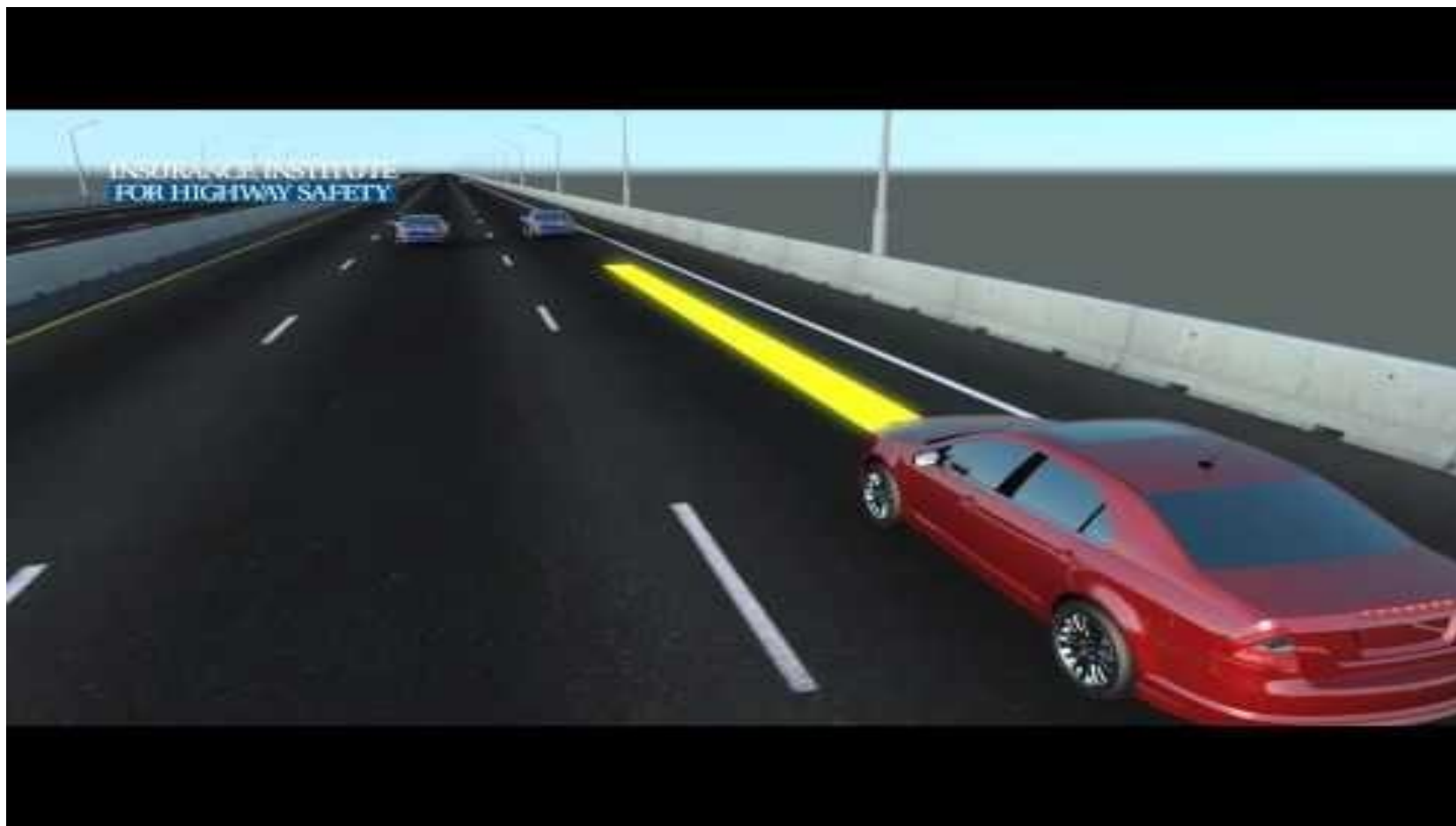
NAMLA Autonomous Omni-Directional Wheeled (i.e. Mecanum Wheeled) Mobile Robot



Source: <https://www.youtube.com/watch?v=xL8deJDusns>



Acceleration Assistance: Adaptive Cruise Control



Source: <https://www.youtube.com/watch?v=GInSPWZRFRM>



Steering Assistance: Lane Keeping Assist



Source: <https://www.youtube.com/watch?v=OQkdvi55woA>



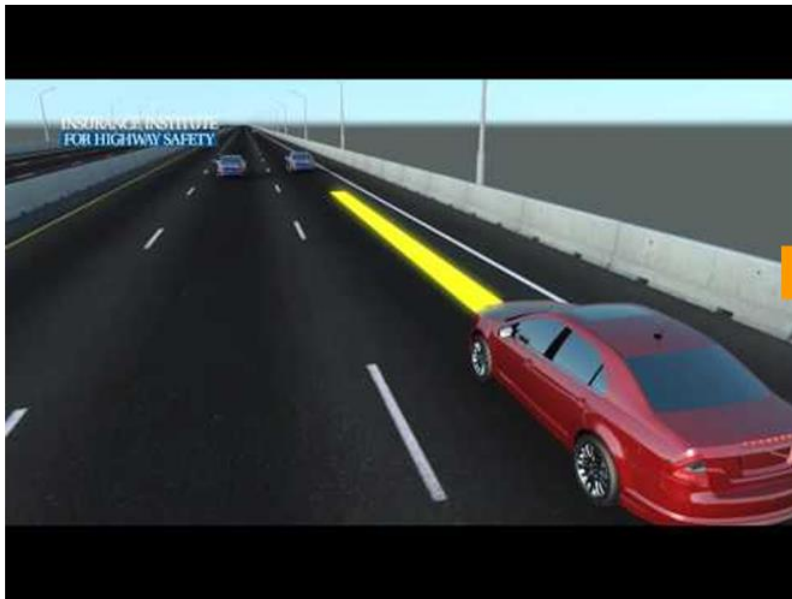
LEVEL 2: PARTIAL AUTOMATION

- STEERING + ACCELERATION
- HUMAN MONITORS ALL TASKS



Combination of Acceleration and Steering

Adaptive Cruise Control



Lane Keeping Assist





SAE Level 2 Fatal Accident



5 fatal
accident (4 in
US, 1 in CN)

All are driver
fatalities

Source: https://www.youtube.com/watch?v=CgLE_ZLLaxw



LEVEL 3: CONDITIONAL AUTOMATION

- MOST DRIVING TASKS
- HUMAN MONITORS ALL TASK



Smart Summon



Source: <https://www.youtube.com/watch?v=nICQG2rg4sw>



Smart Summon (Cont)



Source: <https://www.youtube.com/watch?v=3o2sl37xwOc>



SAE Level 3 Fatal Accident



1 fatal
accident in
2018

Pedestrian
fatality

Source: <https://www.youtube.com/watch?v=ou8sqWr6mF0>

LEVEL 4: HIGH AUTOMATION

- ALL DRIVING TASKS UNDER SPECIFIC CIRCUMSTANCES
- GEOFENCING IS REQUIRED

LEVEL 5: FULL AUTOMATION

- ALL DRIVING TASKS UNDER ALL CONDITIONS

*A geofence is a virtual perimeter for a real-world geographic area; A predefined set of boundaries



Minor Accident in SG



The self-driving car was changing lanes in Biopolis Drive at one-north when it knocked into the lorry. The car was travelling at a "low speed" at the time of the accident --- 2016

What if


























SAE International: Levels of Automation



For on-road vehicles

		 Human driver	 Automated system		
		Steering and acceleration/ deceleration	Monitoring of driving environment	Fallback when automation fails	Automated system is in control
Human driver monitors the road	0 NO AUTOMATION				N/A
	1 DRIVER ASSISTANCE				SOME DRIVING MODES
	2 PARTIAL AUTOMATION				SOME DRIVING MODES
Automated driving system monitors the road	3 CONDITIONAL AUTOMATION				SOME DRIVING MODES
	4 HIGH AUTOMATION				SOME DRIVING MODES
	5 FULL AUTOMATION				

- It is a complex interaction between human drivers and SAE Levels 2 and 3 autonomous vehicles
- Any automated system that removes the human from the driving task, yet requires the human to monitor and supervise the system and regain control when necessary, could be unsafe



‘Ironies of automation’



- **Irony:** combination of circumstances, the result of which is the direct opposite of what might be expected
- *“The mere fact that you can automate does not mean that you should”*
- Humans may **misuse**, **disuse** and **abuse** automation technology
- Humans tend to be **poor supervisors** of automation



‘Ironies of automation’



Source: <https://www.youtube.com/watch?v=kzBNdssqrEA>



Challenges



- **Driver inattention and distraction**
- **Situational awareness**
- **Overreliance and trust**
- **Skill degradation**
- **Motion sickness**



The Future???



- **Re-engaging the driver**
- **The user interface and the communication of automation limitations**
- **Automation misuse and the need to monitor the driver**
- **The personalization of automation**
- **Acceptance**



The Future???



THE STRAITS TIMES

WORLD

Tesla crash victim had lauded 'full self-driving' in videos on TikTok



Since 2016, at least three Tesla vehicles operating on Autopilot have been in fatal crashes. PHOTO: REUTERS



The Future???



Source: <https://www.youtube.com/watch?v=Hl23Yiy-EAE>

From 16:40



HUMAN FACTORS IN LEVEL 5 AUTOMATION

What if





traffic agents: pedestrians, bicycles, cars, buses, etc..





Problem Statement



predict:
future positions of N agents for t_{pred} steps



given:

- history positions
- types (pedestrian, bicycle, car, etc.)
- positions of obstacles



Challenges of predicting their motions

Diverse dynamics, geometry, behaviors
(heterogeneous)





Challenges of predicting their motions

Diverse dynamics, geometry, behaviors
(heterogeneous)

Intensive interactions





Challenges of predicting their motions

Diverse dynamics, geometry, behaviors
(heterogeneous)

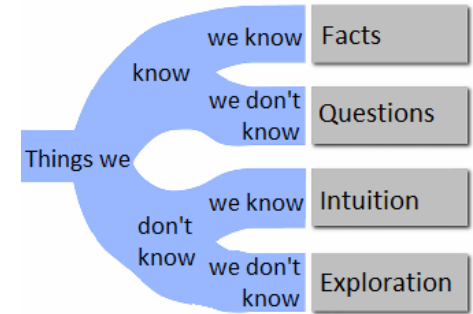
Intensive interactions

Complex road conditions





How real-world agents move?



Traffic agents try to reach their destinations under their kinematic constraints, and in the meantime, they share the responsibility for collision avoidance with each other, while having limited attention capabilities.

The text is annotated with dashed boxes and labels:

- intention** (red dashed box around "destinations")
- kinematics** (orange dashed box around "kinematic constraints")
- responsibility** (red dashed box around "responsibility")
- collision avoidance geometry** (orange dashed box around "collision avoidance")
- attention** (red dashed box around "attention capabilities")

We know these are unknown (uncertain)

We know these are known (quite certain)



Example unknown known: driving is reactive?!

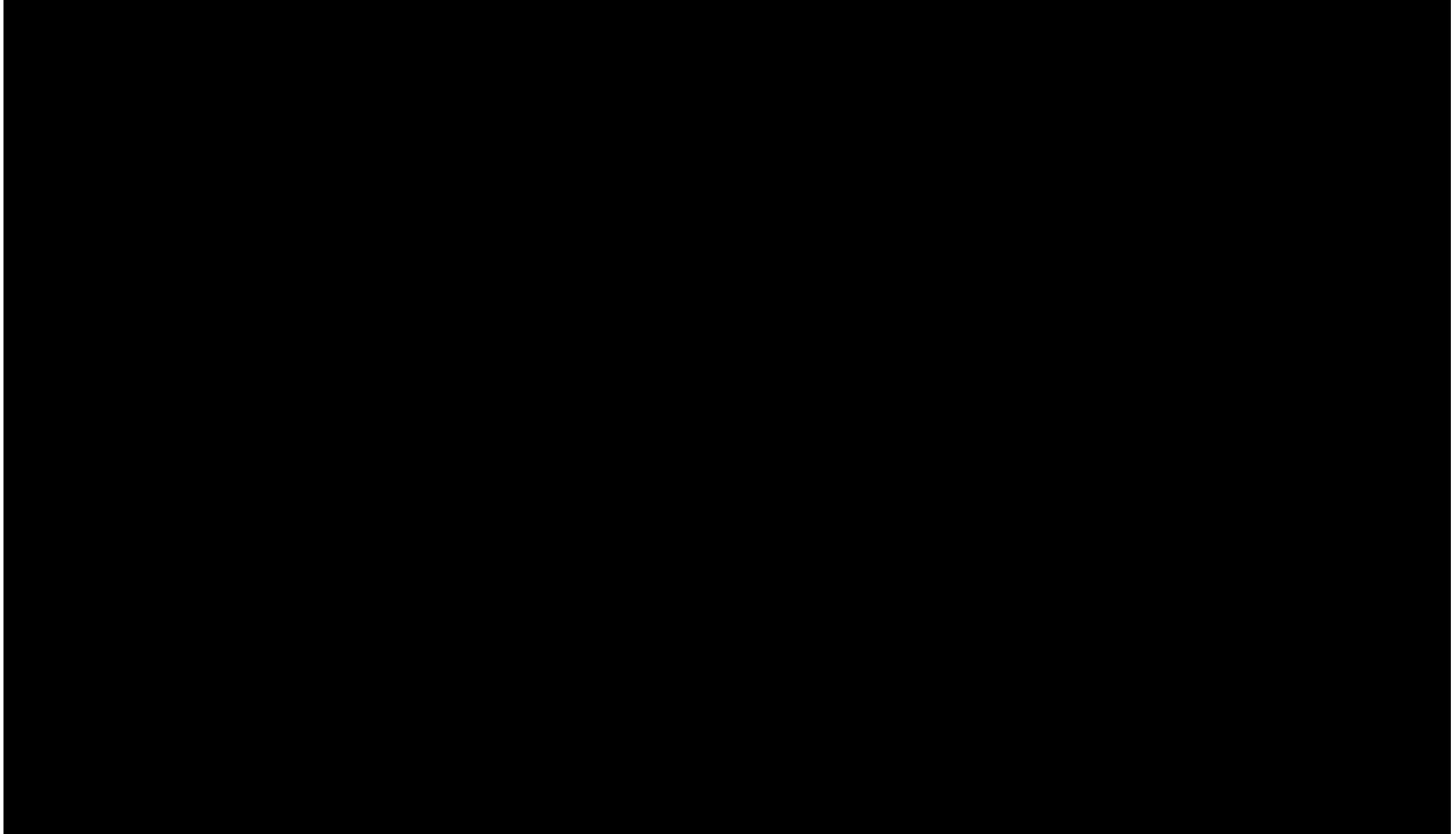
To overtake, the car needs to:

- Understand the stopped car's **intention** of not moving
- Understand the road **context** information such as left/right lane exist and is of same direction





Planning vs Reactive



Source: <https://www.youtube.com/watch?v=SVPVTwxNRLc>



MODULE 3 & 4 WORKSHOP: WORKSHOP DAY 2

WORKSHOP: HANDS-ON CONSTRUCTION
OF MDP/HRIMDP MODELS OF AUTONOMOUS
DRIVING



Personal Healthcare Robot



- Personal assistant robots gain their popularity, E.g., Google Home, Amazon Echo, Xiao Mi robot vacuum, ASUS Zenbo, etc
- They are perfect candidates to become the future personal healthcare givers that monitor the elderly users at the nursing home or in the hospital. In this workshop project, we formulate this interaction between the personal healthcare robot and the human user





Learning Objectives of Workshop Day 2

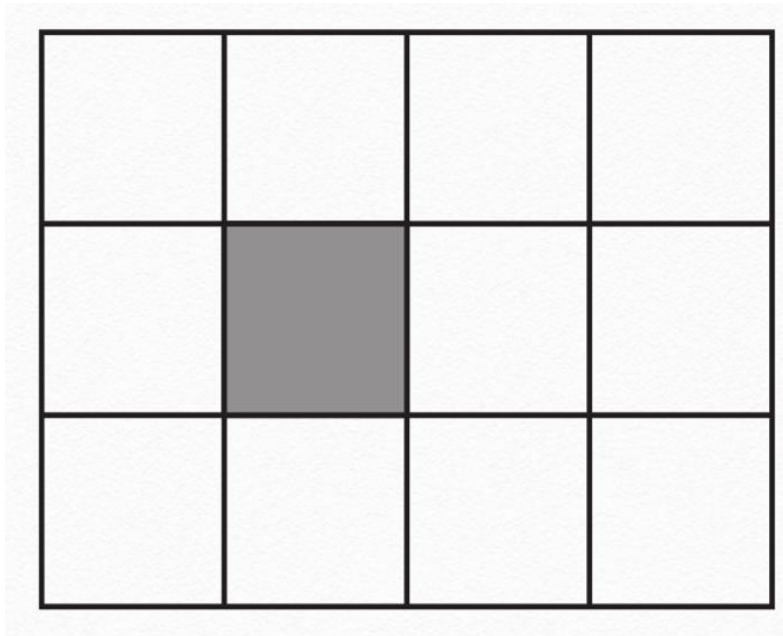


1. Basic programming that creates trajectories of human and robot movements within a virtual world
2. Basic programming that enables you to visualize these interactions between the human and robot (include movements and collision avoidance)
3. Implementation of a Human-Robot-Interaction-focused MDP method (named as HRIMDP) within a 5D gridworld
4. Understand the benefits of the HRIMDP method vs conventional MDP methods

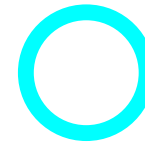
Part A: Introduction to Grid World



Grid World



**The perfect-sized environment
to learn about Reinforcement
Learning for the start**



robot



human

world1.csv

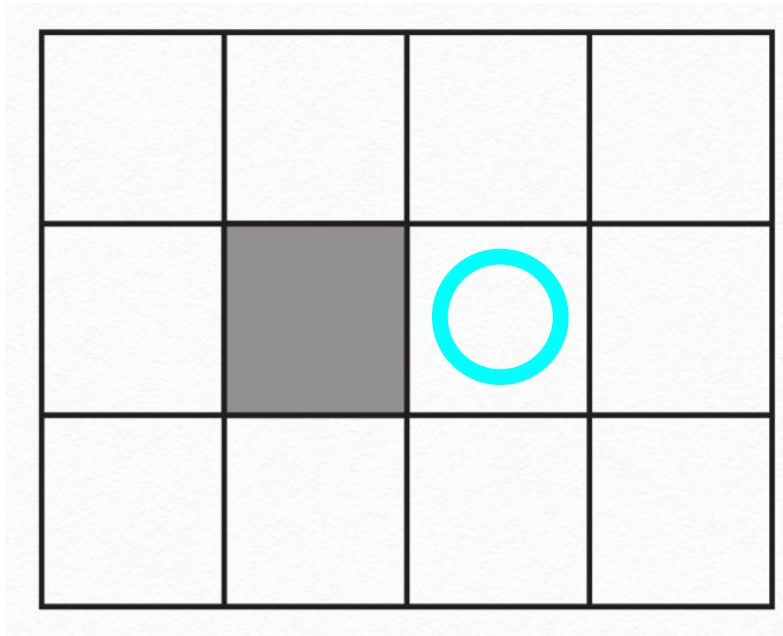
0,0,0,0

0,1,0,0

0,0,0,0



Robot Actions



S: Stay put



U: Go up



D: Go down



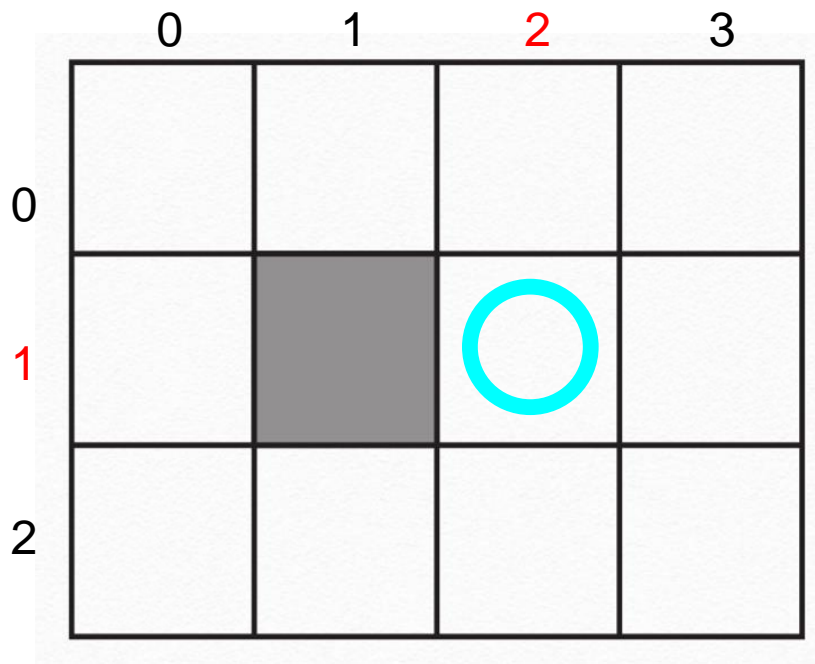
L: Go left (but in this case, stay put **due to collision**)



R: Go right



Robot Execution



robot_a1.csv

U,L,R,R,S,U,L,U,L

> python3 robot_execute.py

world1.csv Gridworld

robot_a1.csv Robot actions

1 2 Robot initial state

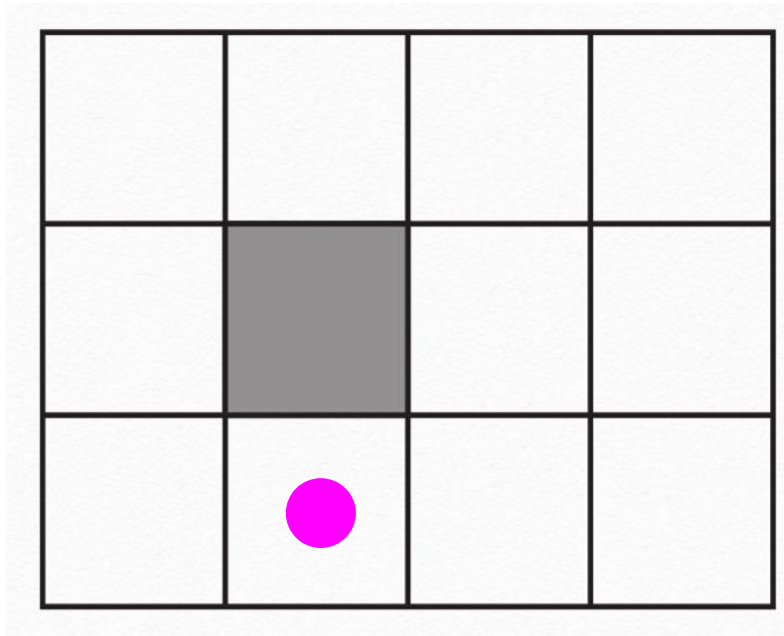
robot_s1.csv Robot trajectory

Executing the sequence of actions results in a robot trajectory:

[[1,2], [0,2], [0,1], [0,2], [0,3],
[0,3], [0,3], [0,2], [0,2], [0,1]]



Human Actions



S: Stay put



U: Go up (but in this case, stay put due to collision)



D: Go down (but in this case, stay put due to collision)



L: Go left



R: Go right



T: Toggle request status



Human Request

-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	10	-1	-1	-1
-1	-1	10	10	10	-1	-1
-1	-1	-1	10	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1

1

-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-5	-1	-1	-1
-1	-1	-5	-10	-5	-1	-1
-1	-5	-10	-10	-10	-5	-1
-1	-1	-5	-10	-5	-1	-1
-1	-1	-1	-5	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1

0

Why the rest of the cells is put at -1 reward instead of 0?

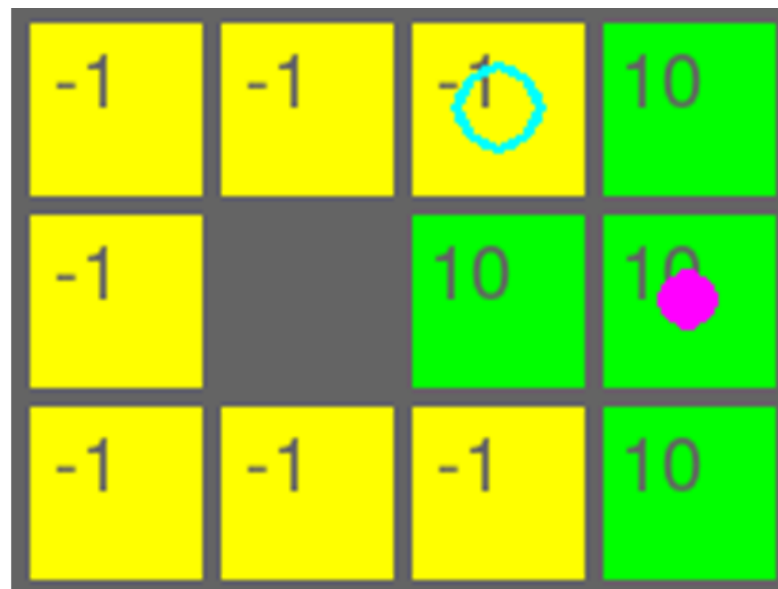


HRI Reward Structure



human_state [0]

“Push” Robot away

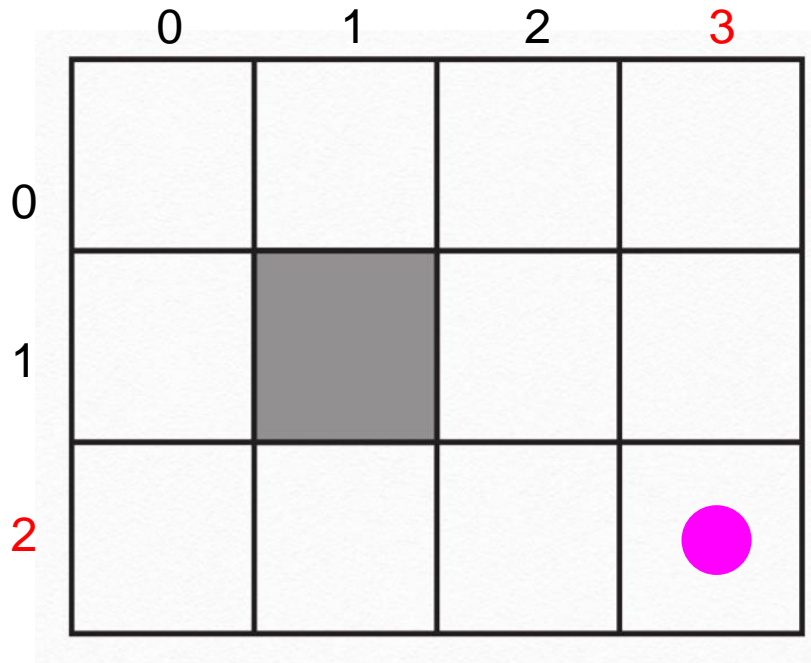


human_state [1]

“Pull” Robot nearer



Human Execution



human_a1.csv

U,L,R,T,D,T,L

> python3 human_execute.py

world1.csv Gridworld

human_a1.csv Human actions

2 3 0 Human initial state

human_s1.csv Human trajectory

Executing the sequence of actions results in a human trajectory:

```
[[2,3],0],[[1,3],0],[[1,2],0],[[1,3],0],  
[[1,3],1],[[2,3],1],[[2,3],0],[[2,2],0]
```



Setup & Preparation



1. In terminal, change directory to the day2a folder

First and last reminder!

2. Run the skeleton code with the following command.

This step is to ensure all the dependencies and packages are properly installed. E.g. python3, pip3, pygame, etc

```
python3 visualizer.py world1.csv human_s1.csv robot_s1.csv
```

3. Install all missing dependencies and packages

Refer to “Setup and Preparation” README file



Complete Run (Try it out!)



1. Generate a valid trajectory of a human based on the init state and a sequence of actions

```
python3 human_execute.py world1.csv human_a1.csv 2 3 0 human_ss1.csv
```

2. Generate a valid trajectory of a robot based on the init state and a sequence of actions

```
python3 robot_execute.py world1.csv robot_a1.csv 0 1 robot_ss1.csv
```

3. Visualize

```
python3 visualizer.py world1.csv human_ss1.csv robot_ss1.csv
```

***Note that you need to install all dependencies beforehand (i.e. python3, pip3, pygame) beforehand**



Clarification



1. The nature of the robot's movements will be dependent on the human's request status
2. Rewards are not collected and factored in yet

Human is visualized as a dot, and the robot is visualized as a circle


-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	10	-1	-1	-1
-1	-1	10	10	10	-1	-1
-1	-1	-1	10	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1

$[[x,y], 1]$

-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-5	-1	-1	-1
-1	-1	-5	-10	-5	-1	-1
-1	-5	-10	-10	-10	-5	-1
-1	-1	-5	-10	-5	-1	-1
-1	-1	-1	-5	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1

$[[x,y], 0]$

DEMO



playback

simulate system dynamics

visualization

Part B: Introduction to HRIMDP

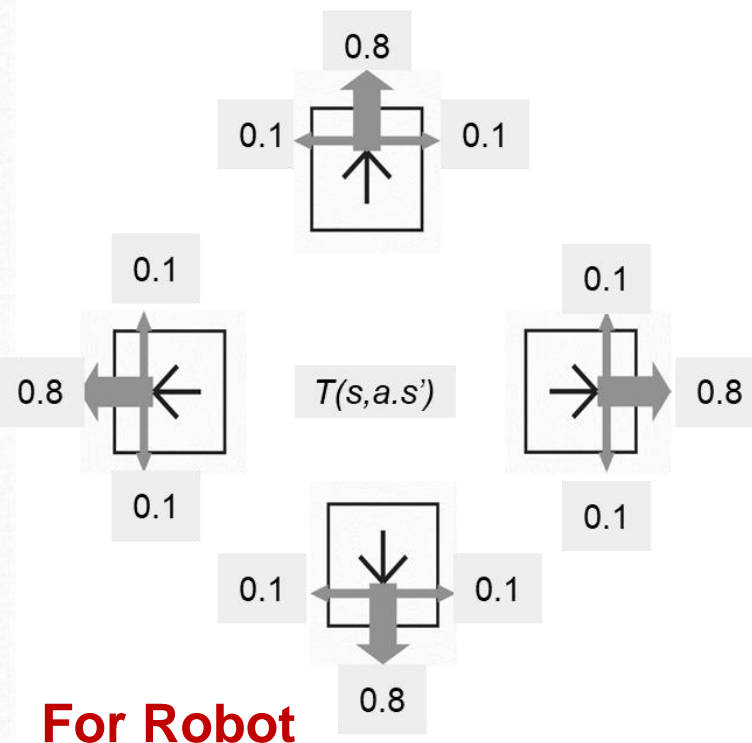
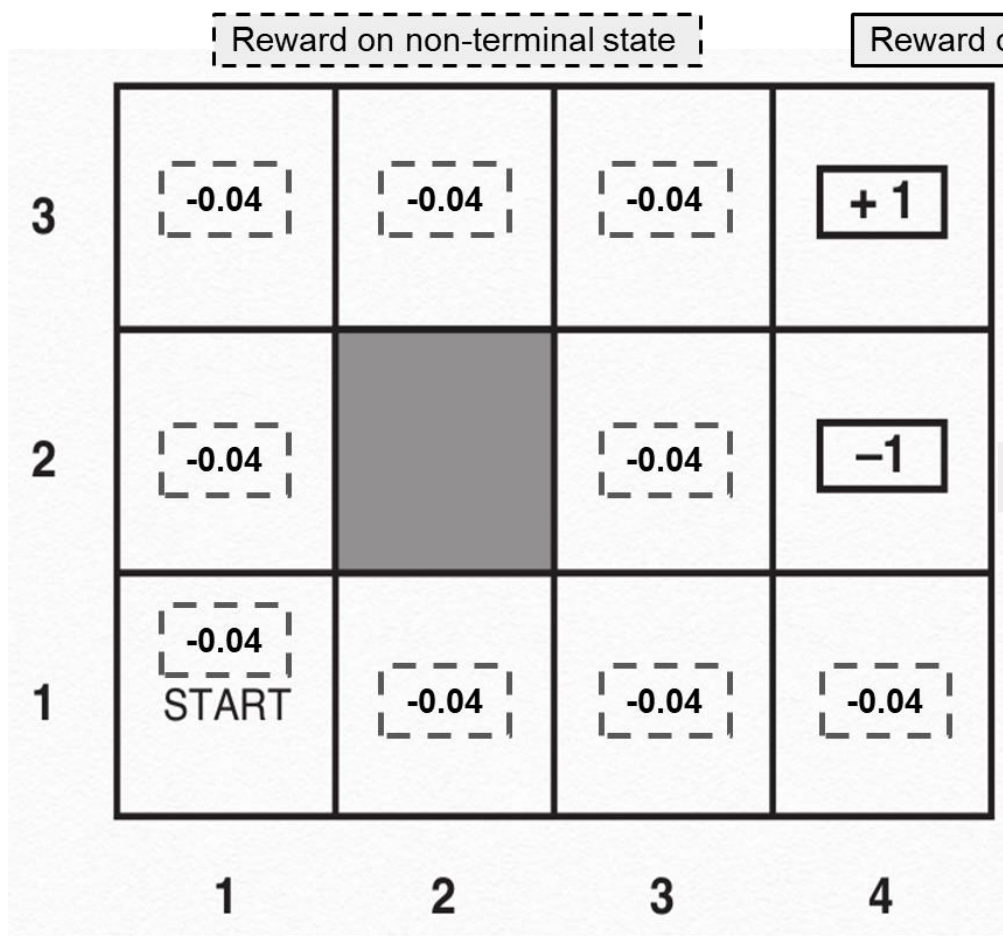


Comparison between GRIDMDP and HRIMDP

	GRIDMDP	HRIMDP
Has HUMAN	NO	YES
State Space	2D robot workspace $S_r = \{robot_x, robot_y\}$	5D joint state space $S_r \times S_h$ $= \{(robot_x, robot_y, human_x, human_y, human_{request})\}$
Action Space	$A_r = \{ 'U', 'L', 'R', 'D' \}$	$A_r = \{ 'S', 'U', 'L', 'R', 'D' \}$
Transition	Uncertain execution	Certain execution But uncertain human state
Reward	Static 2D reward	Dynamic reward in 2D But static in 5D
Terminal States	Yes	No



Reward & Transition Structures used for GRIDMDP



The **reward** function **R** and **transition** function **T** remain unchanged for this example unless stated.



Things to do



1. Verify installation, pytest is needed

```
python3 test_mdp.py
```

2. Try out solving MDP without terminal states

```
python3 test_mdp_without_terminal_states.py
```

3. Implement our HRIMDP with 5 dimensional gridworld

a) Assign **reward[state]**

b) Complete function **calculate_T**

c) Implement function **human_execute_one_step**

4. Test implementation with

```
python3 test_hrimdp.py
```



Hints to complete test_hrimdp.py

1. Assign **reward[state]**

Need a set of codes to represent below reward structure

-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	10	-1	-1	-1
-1	-1	10	10	10	-1	-1
-1	-1	-1	10	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1

[[x,y], 1]

-1	-1	-1	-1	-1	-1	-1
-1	-1	-1	-5	-1	-1	-1
-1	-1	-5	-10	-5	-1	-1
-1	-5	-10	-10	-10	-5	-1
-1	-1	-5	-10	-5	-1	-1
-1	-1	-1	-5	-1	-1	-1
-1	-1	-1	-1	-1	-1	-1

[[x,y], 0]



Hints to complete test_hrimdp.py

2. Complete function **calculate_T**

Assume Transition Matrix of Human Movements is:

		Next Status					
		S	U	D	L	R	T
Current Status	S	0.5	0.1	0.1	0.1	0.1	0.1
	U						
	D						
	L						
	R						
	T						




Hints to complete test_hrimdp.py



3. Complete function **human_execute_one_step**

- Refer to the function before this (i.e. `robot_execute_one_step`) for reference; both should be similar except for:
- You have to account for the following events:
 - a) There is a toggle request from human, upon which, will change the toggle request status. This means if the current status is 'False', this action will change the next status to 'True'; whereas if the current status is 'True', this action will change the next status to 'False'
 - b) There is NO toggle request, upon which, the toggle request status remains the same as the previous

DEMO



playback

simulate system dynamics

visualization



Instructions



This is a group project. Each group submits one zip file of all your codes/files (i.e. py) into LumiNUS at the end of the workshop

A123456_A234567_A345678_P2.zip

- Download all files in the directory **/workshops/day2** for reference codes
- Refer to the README file for instructions



THANK YOU

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