

# Highway simulator: Monitoring update

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Federico Massa, Adriano Fagiolini, Antonio Bicchi, Lucia Pallottino

# Why we needed an update

## What we were doing

- We were trying to monitor the behaviour of a vehicle with a discrete states automaton running different controllers.
- The recognition of the discrete state was made via a forward prediction of the vehicle's continuous state, that was compared with the controller model that we supposed to know (first mayor problem)
- The monitoring itself was achieved by evaluating the rules (that we also supposed to know from the constructor, second mayor problem) that determined the transition between discrete states. Uncertain cases were due to the partial vision of the observer with respect to the monitored vehicle's surroundings.

# Why we needed an update

## What we do now

- We monitor the behaviour of a vehicle based on the abstract features of its movements. This way, we can monitor a vehicle regardless of the dynamic model (it could also be manually-driven)
- Instead of the recognition of the discrete state, we try to recognize the abstract **action** that the vehicle is pursuing.
- The monitoring is now based on the definition of some basic **social rules**, that are defined in an abstract way as the list of conditions that make a certain action forbidden.

# Requirements

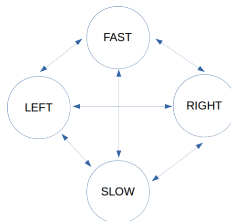
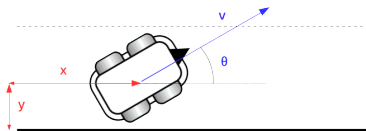
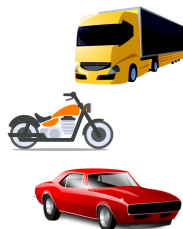
To achieve this, we only need:

- The sensor data of all the vehicles visible by the observer ( $x$ ,  $y$  on the highway plane, at least);
- Some kind of vehicle ID to follow its trajectory;
- A list of possible actions that the monitored vehicle can do;
- A specification of the social rules that are being monitored.

# Configuration

In the configuration file it is possible to choose, for every vehicle in the environment:

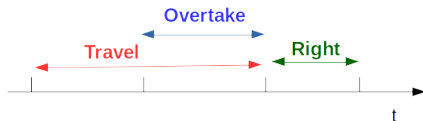
- Initial continuous and discrete state;
- Geometrical parameters (Type of vehicle);
- Dynamic model (Physical Layer type);
- Automaton type;
- Magnitude of sensor errors.



## First step: the action recognition

An action is abstractly defined as a vehicle behaviour in a **finite interval of time**. Examples of it include

- *Travel*
- *Left/Right lane change*
- *Left/Right overtake*
- *Brake*



We can also imagine a more sophisticated action recognition system to include more complicated actions, like *slow left lane change* or *abrupt left lane change*.

### What the monitor needs to know:

The monitor must have a list of possible actions that the observed vehicle can do. The action recognition system is now **independent** from the monitored vehicle's dynamic model.

# How the action recognition works

In order to recognize an action, the monitor needs to:

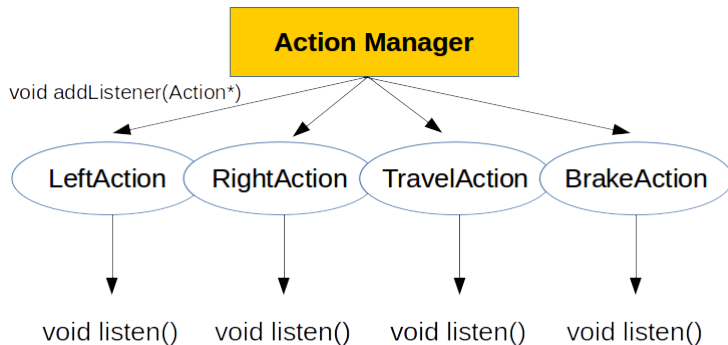
- Identify the observed vehicle;
- Keep track of the visible vehicles' sensor data for a finite interval of time.

Each action is recognized independently  $\rightarrow$  concurrent actions are possible.

An **Action Manager** ( $\Phi$ ) continuously listens to each action ( $a_1, a_2, \dots$ ) and verifies if the trajectory of the monitored vehicle ( $q_1, \dots, q_N$ ) is compatible with that action. If it is, the action is *triggered*.

# The Action Manager structure

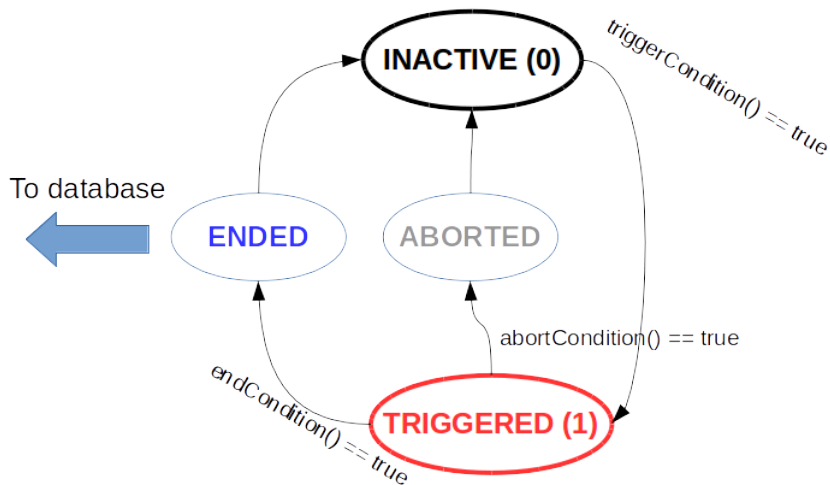
The ActionManager initializes several *listeners*, i.e. actions that the system can recognize.





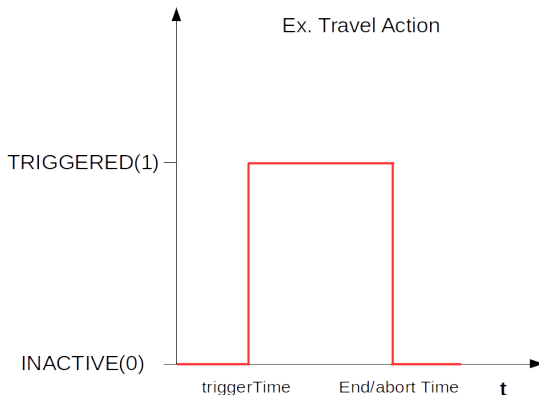
## listen() cycle

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## LeftAction example

We use a N-point trajectory  $(q_1, \dots, q_N)$  and the following parameters:

$\Delta y$ : Transversal distance covered in the recorded trajectory  
 $(y_N - y_1)/N$ ;

$\sigma_y$ : Standard deviation of transversal position.

$f$ : constant between 0 and 1 (needs tuning);

$\Delta y_{tolerance}$ : transversal position tolerance on central position (needs tuning);

$\epsilon_y$ : transverse position sensor error.

**triggerCondition()**  $\Delta y > f \cdot v_{max} \cdot \Delta t_{sim.step}$

$|y_1 - y_{initLane}| < \Delta y_{tolerance} + 3 \cdot \epsilon_y$

**endCondition()**  $|\mu_y - y_{targetLane}| < \Delta y_{tolerance} + 3 \cdot \epsilon_y / \sqrt{N}$

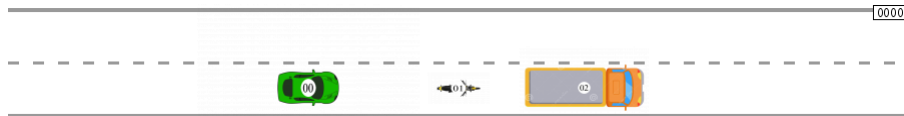
$\sigma_y < \epsilon_y$

**abortCondition()** After trigger time, the vehicle has disappeared.

# LeftAction video



# Complete action example



```

TRAVEL      from 149 to -1  TRIGGERED
RIGHT       from 123 to 148 ENDED
TRAVEL      from 90 to 130  ENDED
LEFT_OVERTAKE      from 41 to 91  ABORTED
TRAVEL      from 28 to 91  ABORTED
LEFT from 1 to 37  ENDED
TRAVEL      from 0 to 7   ENDED
  
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# Complete action example



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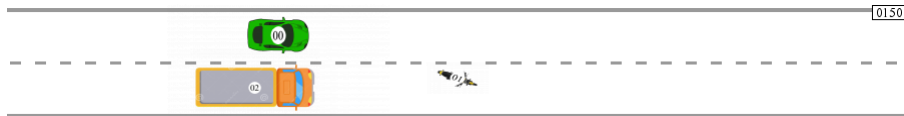
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**Eval mode** Specifies when the rule must be evaluated (when the action is triggered or continuously).

## Example: Italian-like highway rules

### Safety rules

- Safety distance, 1 event: “Someone is in front of the observed vehicle at less than  $x$  meters”;
- Maximum speed, 1 event: “The observed vehicle is faster than  $x$  km/h”

### Left lane change rules

- 1 event: “Nobody is in front of the observed vehicle”;
- 1 event: “Left lane is occupied”;
- 1 event: “The observed vehicle is on the maximum lane”.

### Right lane change rules

- 1 event: “Right lane is occupied”;
- 1 event: “The observed vehicle is on the minimum lane”.

### Travel rules

- 1 event: “Right lane is free”;

**Left overtake** allowed, **Right overtake** forbidden.



# Action Manager and Rule Monitor

An object called **Rule Monitor** is used to check if the rules are violated, as the Action Manager “listens” to each action.

These two objects are independent, but they can communicate via the **rules categories**.

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- Keeps the action and the rule systems well separated;
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- Keeps the action and the rule systems well separated;
- Different actions can share some rule categories.

→ This also allows to make complex action managers without changing the rule system.

# Action and rules

Action	Rule categories
LeftAction	Safety Left Lane Change
RightAction	Safety Right Lane Change
TravelAction	Safety Cruise
LeftOvertakeAction	Safety Cruise
RightOvertakeAction	Safety Right Overtake

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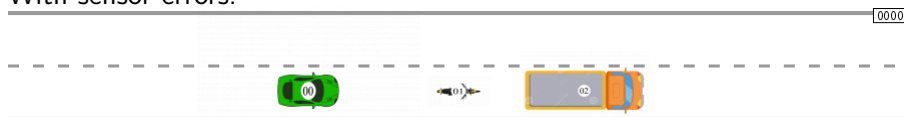
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4. If the Evaluation Mode is on TRIGGER, the rule is checked only at trigger time. If it is on CONTINUOUS, the rule is checked every time until the action ends (e.g. Safety rules are generally in CONTINUOUS mode);
5. A rule is **True** if a wrong behaviour was spotted (at least one of the events is true);  
**Uncertain** if the observer cannot tell if a rule is verified or not due to hidden areas (no events are true, at least one uncertain);  
**False** if the behaviour is correct (all events are false).

# LeftAction example

Without sensor errors:



With sensor errors:



## Remarks on the C++ code

The C++ code is mostly written using polymorphism, so that it is very easy to make a more complex system without having to understand the underlying code:

- Specific actions inherit from the Action class. To add a new action, it is only necessary to manually write the trigger, end and abort conditions.
- Specific set of rules inherit from the SocialRules class. After manually writing a new set of rules it is possible to change the monitored rules by changing a single line of code.

A similar concept applies when wanting to change the vehicle geometry, dynamics or automaton. The code is written keeping flexibility in high regard.

## Summing up

The simulator should now be able to:

- Recognize and record list of actions done by the monitored vehicles;
- Verify, if possible, if the monitored vehicle is following the rules;
- Do this no matter the vehicle size, dynamic model, state transition rules (could also be manually driven in theory).

### Critical point:

Action parameters tuning

Further studies:

- Use of consensus to improve the observer dataset (both to reduce sensor errors and to reveal the content of the hidden areas);
- Reputation system: how to deal with infractions?
- Roomba?
- Possible application in Roborace?
- Besides highways?