## Dream-like simulation abilities for automated cars



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1.9	27/June/2018	Mauro Da Lio	Final version including minor edits from Gas- tone Pietro Rosati Papini (mostly typos).

## **Executive Summary**

Work Package 3 spans from month 12 to month 30. The current deliverable (D3.1) is a checkpoint, at month 18, to verify that efficient methods have been developed for dreaming (to be used in project phase 2, from month 19 to 30).

We think that we have developed a spectrum of methods that cover all aspects of the dreaming mechanism, though not all the methods presented here are equally developed at this stage.

Some approaches have produced results that go beyond the hopes (and even beyond the DoA). These include the episodic simulation machinery of sections 3 and 4 and the embodied simulation mechanisms of section 6.

We have indeed identified and developed two distinct imagery mechanisms.

The first one is referred to as *"episodic simulations"* and is related to cortical-subcortical processes in the brain. We have developed two distinct types of implementations for this mechanism.

- The former (section 3) relies on artificial neural networks that reproduce convergent divergent zones (CDz) posited to form the dorsal stream. We have significantly demonstrated and tested this mechanism in section 3, with reference to imagery of roads and imagery of other road users' behaviours.
- The second mechanism for episodic simulations (section 4) is less tied to biological inspiration but very effective and complementary. It is based on Genetic Algorithm (GA) that swaps and recombines the elements of the scenario and that can be used to set up simulation environments. We have extensively tested this mechanism with one scenario on curvy roads and another inspired by the recent UBER fatality. Furthermore, GA algorithms are potentially effective in directing the dreaming exploration towards those situations that are likely to create useful dreams (but this has not been tested yet): accordingly, the GA machinery acts as a system that is adversarial to the learning agent in a way that, while the agent improves, the dreaming machine creates more challenging situations.
- The two mechanisms here described can act in synergy, where the GA is responsible of creating discrete elements in the scenarios and the CDz model continuous items such as, e.g., other road users.

The second imagery mechanisms that have been developed are the so-called *"embodied simulation mecha-nisms"* which are related to the cerebellar loop (section 6).

- In sections 6.1 and 6.2 we have shown effective ways to learn *forward models* either based on carefully engineered neural networks that resembles the cerebellar adaptive filters (section 6.1) or based on more traditional control theory and parametric modelling (section 6.2).
- We then show how to derive inverse models in section 6.3 and how to use them for vehicle low-level motor control (which is important for both on the fly adaptation and interoperability).
- Finally, in section 6.4 we present a method for learning (bootstrapping) a hierarchical motor control architecture that steps from forward models to motor primitives to sequences of actions (i.e., a sub-sumption architecture). This has the potential for discovering and optimizing behaviours made of sequences of actions (further work is required here but a demonstration is already given in section 6.4).

Concerning learning of safe behaviours, section 5 presents an extension to the agent architecture based on the notion of frontal-cortex biasing. Here the basal-ganglia MSPRT loop (action selection mechanism based on Multi-Hypothesis Sequential Probability Ratio Test) is extended with biasing matrices that represent choices corresponding to longer-term intentions.

- The Logical Reasoning Module (LRM) implements the traffic regulations (Highway code) in a subsumptive PA (perception-action) format and is responsible (together with the main dorsal stream) for producing the various potential biases in intention.
- An algorithm implements Reinforcement learning at action-selection level, ideally allowing the agent to learn how to suppress immediately rewarding actions in favour of better longer terms rewards.

These algorithms have been coded, but extensive testing remains to be done.

While the LRM can initiate exploratory learning across the PA hierarchy as a whole (as per the project specification), a potentially limiting issue that we have realized recently concerns the fact that there is currently relatively little scope for explicit learning in the LRM. Another limitation that has emerged is related to practical limitations in the Reinforcement Learning algorithm itself at the action-selection level, where for practical reasons the number of alternatives has to be limited and this limits the learning capacity of the agent with this respect.

On the other hand, these limitations are overcome by the action sequence discovery mechanism described in section 6.4. So, while the LRM is useful to embed traffic rules (that are coded) the bootstrapping machine in section 6.4 allows to discover behaviours that might be useful in emergency situations.

Finally, concerning interactions of the frontal stream with the dreaming mechanisms themselves (section 4), the LRM is also designed to act in a generative manner, hallucinating high-level legal road configurations (and corresponding legal intentions) for the GA dreaming mechanism. Following the specification and development of the GA system, it has been determined that the LRM interface for dreaming will be required to occur at a higher level than that of the run-time system, and thus require a separate interfacing protocol at a higher level of PA subsumption. This part is under development.