

Autonomous Decision-Making with Incomplete Information and Safety Rules based on Non-Monotonic Reasoning

José-Luis Vilchis-Medina, Charles Lesire-Cabaniols^a,
Karen Godary-Dejean^b

ONERA – The French Aerospace Lab^a
LIRMM, Université de Montpellier, CNRS^b

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Overview

- 1 Introduction
- 2 State of the art
- 3 Non-monotonic Reasoning
- 4 Contribution
- 5 Simulation and Results
- 6 Conclusion

Description of the robot

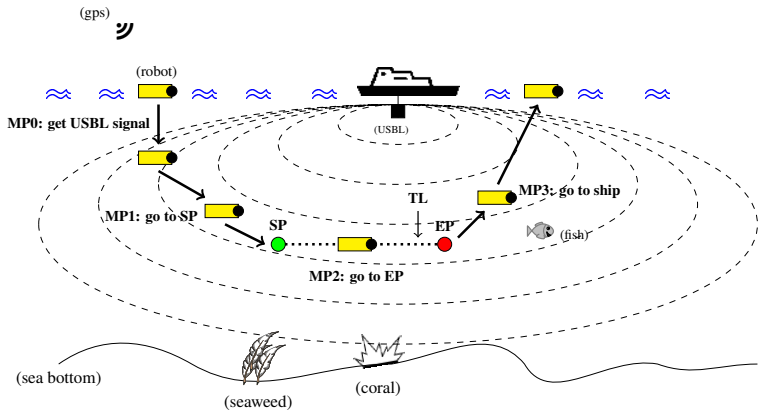
Underwater robot

- dimension 93cm L, 41cm W, 23cm H,
- micro-computer Intel Atom Z8000 processor,
- 4 batteries, 8 motors, weight ≈ 30 Kgs



Sensor name	ID	Protocol
Leak indicator	SOS-Leak-Sensor	Digital
Flasher	Led	Digital
Bar30	MS5837	i2c
IMU	BNO055	i2c
Camera	-	IP
GPS-Robot	DP0104	uart
Echo-sounder	Ping Sonar	uart
SeaTrac	X150 USBL Transponder	uart

Example of a normal mission : "transect"



Problematic

Generally, autonomous robots are confronted with unexpected situations due to multiples causes:

- changes in the environment,
- uncertain information,
- failures, etc.

Example (in our case):

temperature increase, water leakage, non geo-referenced location, exceeded depth, mission overrun time, drifting from transect line, etc.

Solution/Proposition

Goal reasoning: it is to decide on current information, while always considering the objective(s) and safety of the robot.

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State of the art

- KnowRob framework¹:
 - vague information treatment to perform tasks: “set the table”, “clean up” . . .
 - Description Logic for knowledge representation,
 - central knowledge consultation via Prolog (classical proving theorems)
- Answer-Set Programming (ASP):
 - declarative programming and stable model paradigm²,
 - generally has difficulties to reason on all classes of stable models³,
 - does not support free variables and difficult to debug a program
- I proposed (on my Ph.D. thesis):
 - a formalization for modeling a solar glider’s piloting behavior using a non-monotonic logic,
 - the use of Default Logic (DL) and Prolog (non-monotonic reasoner),
 - a framework for the study of resilient systems using DL.

¹Tenorth, Moritz, and Michael Beetz. “KnowRob: A knowledge processing infrastructure for cognition-enabled robots.” (2013).

²Erdem, Esra, Erdi Aker, and Volkan Patoglu. “Answer set programming for collaborative housekeeping robotics: representation, reasoning, and execution.” *Intelligent Service Robotics* 5.4 (2012): 275-291.

³Elkhatib, Omar, Enrico Pontelli, and Tran Cao Son. “ASP — PROLOG: A System for Reasoning about Answer Set Programs in Prolog.” *International Symposium on Practical Aspects of Declarative Languages*. Springer, Berlin, Heidelberg, 2004.

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Non-monotonic Reasoning

The most important proposal by:

J. McCarthy (*Circumscription '80*), **R. Reiter** (*Default logic '80*)...

- New information can invalidate previous conclusions,
- Resolve contradictions,
- Reasoning about knowledge,
- Rational conclusions from partial information.

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Formally, monotonicity:

$$A \rightarrow \omega$$
$$A \cup B \rightarrow \omega$$

Default Logic [Reiter]

Definition

A default theory is a pair $\Delta = (D, W)$, where D is a set of defaults and W is a set of formulas in FOL.

- A default d is: $\frac{A(X):B(X)}{C(X)}$
- $A(X), B(X), C(X)$ are well-formed formulas
- $X = (x_1, x_2, x_3, \dots, x_n)$ is a vector of free variables(non-quantified).

Intuitively a default means, **“if $A(X)$ is true, and there is no evidence that $B(X)$ might be false, then $C(X)$ can be true”**.

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“All birds fly”, $\forall X, bird(X) \rightarrow fly(X)$ (chickens, penguins, kiwis...!?)

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$$\forall X, bird(X) \wedge \neg chicken(X) \vee \neg penguin(X) \vee \dots \rightarrow fly(X)$$

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$$\forall X, bird(X) \wedge \neg chicken(X) \vee \neg penguin(X) \vee \dots \rightarrow fly(X)$$

Example (Default logic):

“Normally, the birds fly”, $D = \frac{bird(X):fly(X)}{fly(X)}$

$$W = \{bird(tweety), penguin(tweety) \rightarrow bird(X), penguin(X) \rightarrow \neg fly(X)\}$$

Default Logic [Reiter]

Definition

E^Δ is an extension of Δ iff:

- $E_0 = W$ and
- for $i > 0$, $E_{i+1} = Th(E_i) \cup \{C(X) \mid \frac{A(X):B(X)}{C(X)} \in D, A(X) \in E_i \wedge \neg B(X) \notin E^\Delta\}$
- $E^\Delta = \bigcup_{i=0}^{\infty} E_i$

Property

If every default of D is normal: $\frac{A(X):C(X)}{C(X)}$

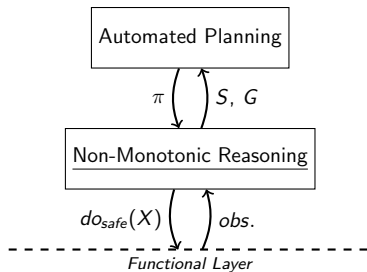
$\neg B \notin E^\Delta$ is replaced by $\neg C \notin E_i$

there is always one extension and help to perform greedy algorithm

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Contribution



Automated Plan.: FF algorithm

Non-Monotonic Reasoning model

$\Delta = (D, W)$ where:

$$D = D_{est} \cup D_{safety} \cup D_{goal}$$

$$W = W_{obs} \cup W_{est} \cup W_{safety} \cup D_{goal}$$

Functional layer: representation of the robot capabilities and features (*skill model*) [LesireDG20]

Notation

observations: $obs.$; **State and Goal:** S, G ; **Sequences of actions:** π ;
safe actions applied: $do_{safe}(X)$

Guidelines of the proposed model

Observations:

From *skill model* (resources states, skill execution statuses)

$$W_{obs} = \{at(home) \wedge \neg usbl_captured \wedge gps_captured \wedge on_surface \wedge \dots\}$$

Estimation theory:

Extended knowledge via NMR (non-obvious information)

- $\Delta_{est} = (D_{est}, W_{est})$
 - $d_{loc} = \frac{T:localized}{localized}$ (localized by default)
 - $\varphi_{loc} = \neg usbl_was_captured \wedge \neg gps_was_captured \rightarrow \neg localized$
 - $\varphi_{loc'} = low_loc_precision \rightarrow \neg localized$

$$d_{loc} \in D_{est}$$

$$\{\varphi_{loc}, \varphi_{loc'}\} \in W_{est}$$

Guidelines of the proposed model

Safety theory:

Emergency rules (**critical failure**: temperature increase, water leakage. . .)

- $\Delta_{safety} = (D_{safety}, W_{safety})$

- $\frac{A(X):do_{safe}(a)}{do_{safe}(a)}$

$$d_{safe} = \frac{safety_sensor_failure(X) : do_{safe}(shut_down())}{do_{safe}(shut_down())}$$

$$d_{safe'} = \frac{next_action(a) : do_{safe}(a)}{do_{safe}(a)}$$

- $A(X) \rightarrow do_{safe}(a)$ or $A(X) \rightarrow \neg do_{safe}(a)$

$$\varphi_{safe} = low_energy \rightarrow do_{safe}(shut_down())$$

$$\varphi_{safe'} = \neg localized \wedge next_action(transect(X, Y)) \rightarrow \neg do_{safe}(transect(X, Y))$$

$$\{d_{safe}, d_{safe'}\} \in D_{safety}$$

$$\{\varphi_{safe}, \varphi_{safe'}\} \in W_{safety}$$

Guidelines of the proposed model

Goal theory:

Deriving the current mission objective

- $\Delta_{goal} = (D_{goal}, W_{goal})$
 - $\frac{A(X):goal(Y)}{goal(Y)}$ or $\frac{A(X):\neg goal(Y)}{\neg goal(Y)}$

$$d_{goal} = \frac{\top : goal(transect_done(p_A, p_B))}{goal(transect_done(p_A, p_B))}$$

$$d_{goal'} = \frac{\neg enough_energy(transect(p_A, p_B)) : \neg goal(transect_done(p_A, p_B))}{\neg goal(transect_done(p_A, p_B))}$$

- $A(X) \rightarrow goal(Y)$ or $A(X) \rightarrow \neg goal(Y)$

$$\varphi_{goal} = \neg localized \rightarrow \neg goal(transect_done(X, Y))$$

$$\varphi_{goal'} = close_seabed \rightarrow goal(on_surface)$$

$$\{d_{goal}, d_{goal'}\} \in D_{goal}$$

$$\{\varphi_{goal}, \varphi_{goal'}\} \in W_{goal}$$

Guidelines of the proposed model

Programming syntax used

```
%----- Language
% on_surface           : remi is on surface (T) or in water (F)
% low_loc_precision   : localization precision is too low
% low_loc_precision_transect: localization precision is too low for transect
% timeout_task(X)     : X an action; action X has just timed out
%                     : (then aborted by the functional layer)
% low_energy          : energy under a safety threshold
% enough_energy(X)    : X an action; energy is sufficient to perform X
% collision            : collision (T / F)
% detected(X)         : specie X has been detected
% robot_task(X)       : X an action; robot is performing action X
% next_task(X)        : X an action; the planner has returned X as next action
% transect_done(X, Y) : X, Y start/end locations
% at(X)               : X robot location
% timeout_mission     : timeout mission (T / F)

%----- ACTIONS
% go_boat
% go_surface
% init_usbl
% transect(X, Y)
% shut_down
% diagnose_motors
```

Guidelines of the proposed model

Programming syntax used

```
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%% (W_safety)
cl('depth_problem',dur,[sensor_problem(depth)], do_safe(shut_down), 100).
cl('controllability',dur,[-control_actuators], do_safe(shut_down), 100). %(safe4)
cl('SW_mods_active',dur,[-sw_mods], do_safe(shut_down), 100). %(safe5)
cl('problem_motors',dur,[-all_motors_ok], do_safe(diagnose_motors), 100).
cl('collision',dur,[collision], do_safe(shut_down), 100). %(safe8)
cl('low_energy_total',dur,[low_energy], do_safe(shut_down), 100).
cl('not_loc_do_other',dur,[-localized, next_task(transect(_,_))],do_safe(shut_down),100).

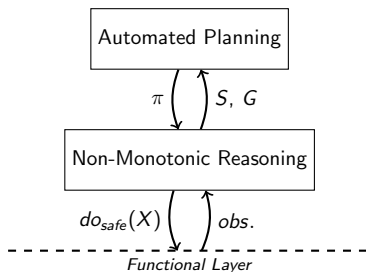
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%% (D_safety)
cl('safety_sensor_problem', def, [safety_sensor_problem(_)], do_safe(shut_down), 100).
cl('do_action', def, [next_task(X)], do_safe(X), 100).

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%% (W_est)
cl('not_geoloc', dur, [-usbl_was_captured, -gps_was_captured], -localized, 0).
cl('precision_min',dur,[low_loc_precision], -localized, 0).

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%% (D_est)
cl('loc_by_def', def, [], localized, 0).

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%% (D_goal)
cl('sensor_problem', def, [sensor_problem(_)], goal(on_surface), 45).
cl('video_problem', def, [sensor_problem(video)], -goal(on_surface), 55).
```

Non-monotonic Reasoning Process



Algorithm NMR process implemented

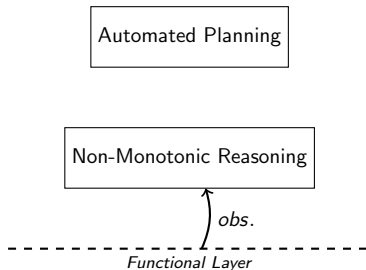
Require: $D, W \notin \emptyset$

Ensure: $mission_done \leftarrow \perp, W_{obs} \leftarrow \emptyset$

```

1: repeat
2:    $W_{obs} \leftarrow obs$ 
3:    $\Delta = (D, W)$ 
4:    $E^\Delta = \{S, G, do_{safe}(a)\}$ 
5:   if  $\exists \{do_{safe}(a)\} \subset E^\Delta$  then
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7:   else if  $\exists \{S, G\} \subset E^\Delta$  then
8:      $\pi \leftarrow AutomatedPlan(S, G)$ 
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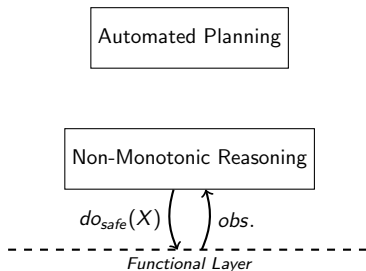
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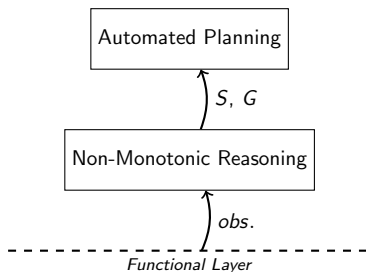
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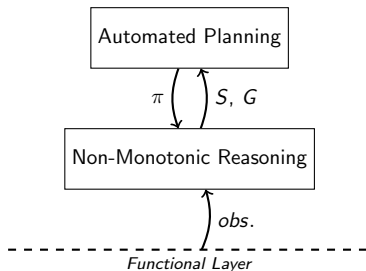
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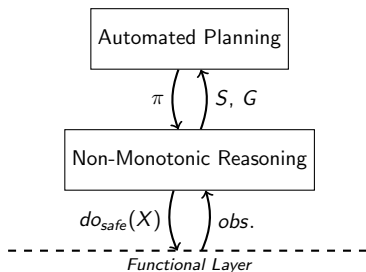
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Experimental Procedure

- **Skills interface**⁴ (model + manager)
 - operational and descriptive model of the robot
 - 3 data, 9 resources and 10 skills or actions
 - ROS2 middleware (Python/Prolog)
- **Automated Plan.** (FF algorithm)
- **Default Theory** (Prolog)
 - behaviors (goals + safety): modeled with 44 rules (17 normal defaults and 27 exceptions)
 - part of the behaviors came from a risk analysis⁵

Evaluation:

Two simulations were performed, calculation time for a simple as well as a complex problem was evaluated.

⁴ Lesire, Charles, David Doose, and Christophe Grand. "Formalization of Robot Skills with Descriptive and Operational Models." IROS, 2020.

⁵ Hereau, A., Godary-Dejean, K., Guiochet, J., & Crestani, D. (2021, May). A Fault Tolerant Control Architecture Based on Fault Trees for an Underwater Robot Executing Transect Missions. In International Conference on Robotics and Automation (ICRA 2021).

Simulation and Results

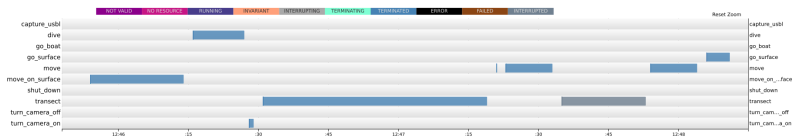


Figure: low localization timeline

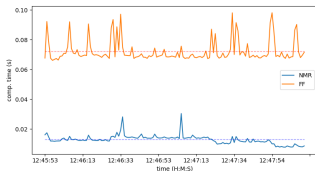


Figure: low localization computation time

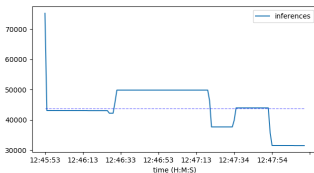


Figure: low localization inferences

Simulation and Results

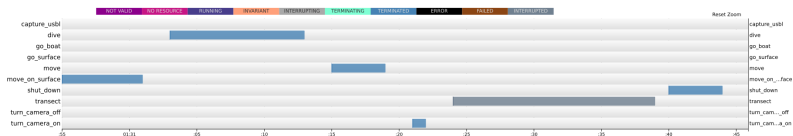


Figure: shutdown timeline

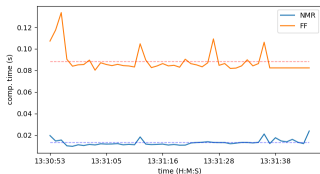


Figure: shutdown computation time

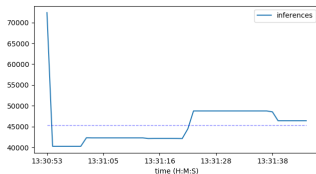


Figure: shutdown inference

Simulation and Results

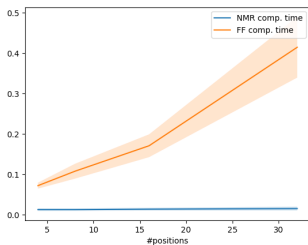


Figure: computation time of 32 positions

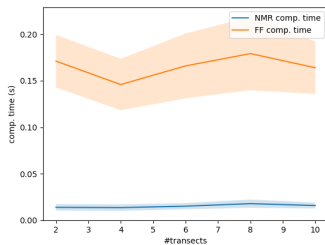


Figure: computation time of goals

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Conclusion

- New decision-making architecture based on a non-monotonic logic:
 - Goal reasoning,
 - Safety rules management
- Default logic is a promising tool for tackling problems that have non-monotonic behavior,
- Model guidelines for use in others applications,
- Practical verification of model complexity (quasi-linear) with 44 rules (17 defaults and 27 exceptions)

Future work:

Multi-agents systems, autonomous agents, other applications. . .

Thank you for your attention,

Questions ?⁶

⁶email: jose.vilchis@ensta-bretagne.fr