

Using Decentralised Conflict-Abduction-Negation in Policy-making

Étienne Houzé^{1,2}, Ada Diaconescu¹, Jean-Louis Dessalles¹

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¹Télécom Paris

²EDF R&D



Problematic cases

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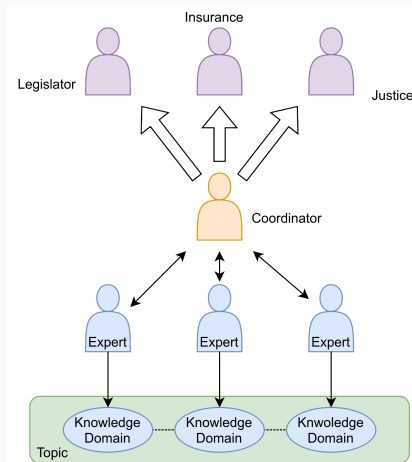
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Can we provide a formal canvas for reasoning in these two cases ?

Problematic cases

Complex problems

In many real-world situations, a policy affects an entire topic, spanning over the knowledge of different experts. As such, to find relevant policies and causal chains, coordination and communication between these experts is required.

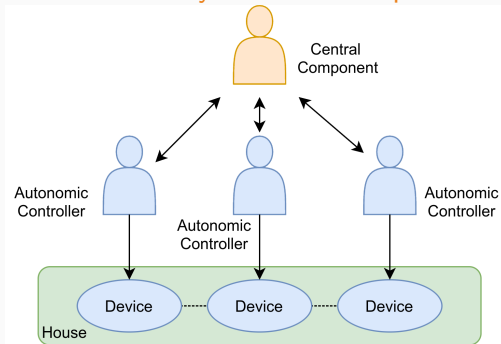


Insights from Smart Homes

What is a Smart Home ?

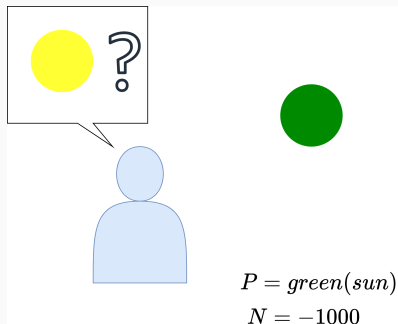
A Smart Home is a collection of connected devices located in the same house acting together towards the completion of various high-level goals, such as comfort, power efficiency, security[3].

Parallel between Smart Home Systems and complex decisionary systems !



The D-CAN process

Conflict – Abduction – Negation [1]



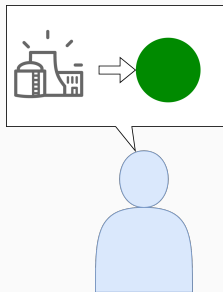
Conflict

At root of the process is a conflict, *i.e.* an proposition P that is realized while not expected/wished. A proposition is assigned a *necessity* N to convey intensity information.

Abduction

Negation

Conflict – Abduction – Negation [1]



$C = \text{polluted}(\text{air})$
 $N = -1000$

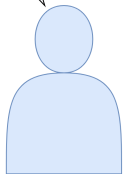
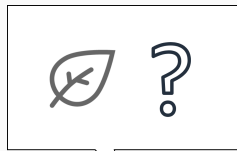
Conflict

Abduction

Observing a conflict (P, N) , we propagate the conflict to a hypothetical cause C , creating the new conflict (C, N) .

Negation

Conflict – Abduction – Negation [1]



$C = \neg polluted(a)$
 $N = 1000$



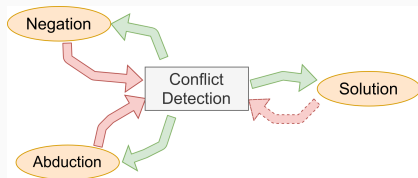
Conflict

Abduction

Negation

When confronted to a conflict (P, N) , we can consider its negation $(\neg P, -N)$ as a conflict of the same intensity, thus propagating the reasoning to other branches.

Conflict – Abduction – Negation [1]



Conflict

Abduction

Negation

We iterate over these steps, propagating the conflict, until a solution is found, or the conflict is given up!

Decentralizing the process[2]

The Spotlight

Generic, no specialized knowledge. Analyzes the request and route it to the relevant component

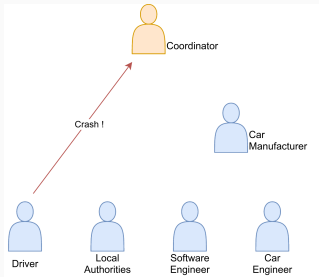
The Explanatory Component

Local Knowledge, specialized. Detects conflicts, uses abductive reasoning, “points finger” at the responsible.

Illustrative Examples

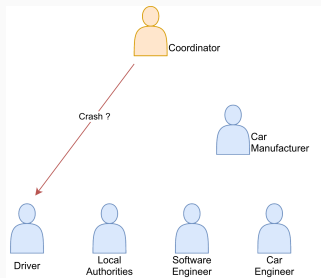
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1. The driver reports her issue : the crash $P = crash, N = -100$



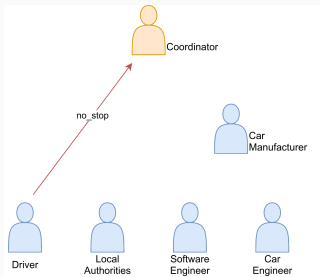
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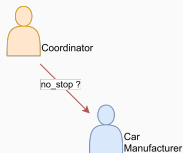
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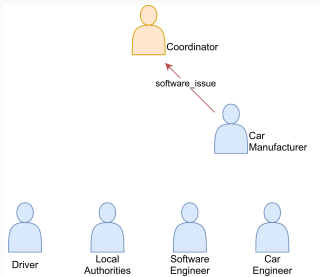


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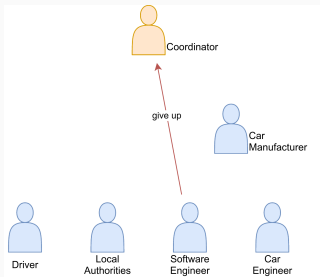


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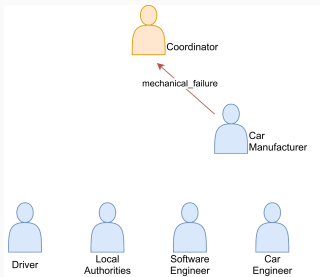
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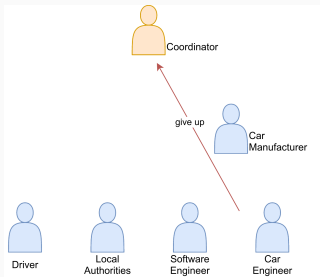
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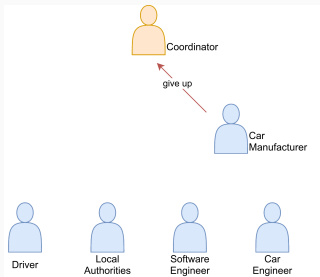
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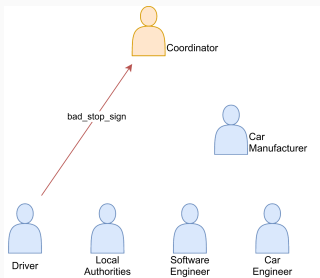
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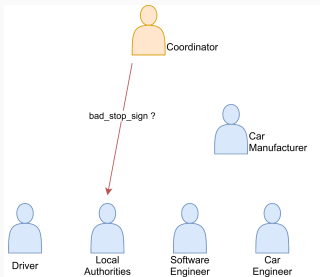
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11. Go to the local authorities...

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Conclusion

- The D-CAN process enables inter-agent communication in Smart Homes.
- This process can be transferred to policy-making situations where different experts must cooperate.
- It provides a formal canvas to enable discussion and model interactions between these experts.
- Limits : Human agents can deceive, make mistakes, hide the truth !

Thank you !

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