

Trace-Diagnostic for Signal Temporal Properties: an Evolutionary Approach

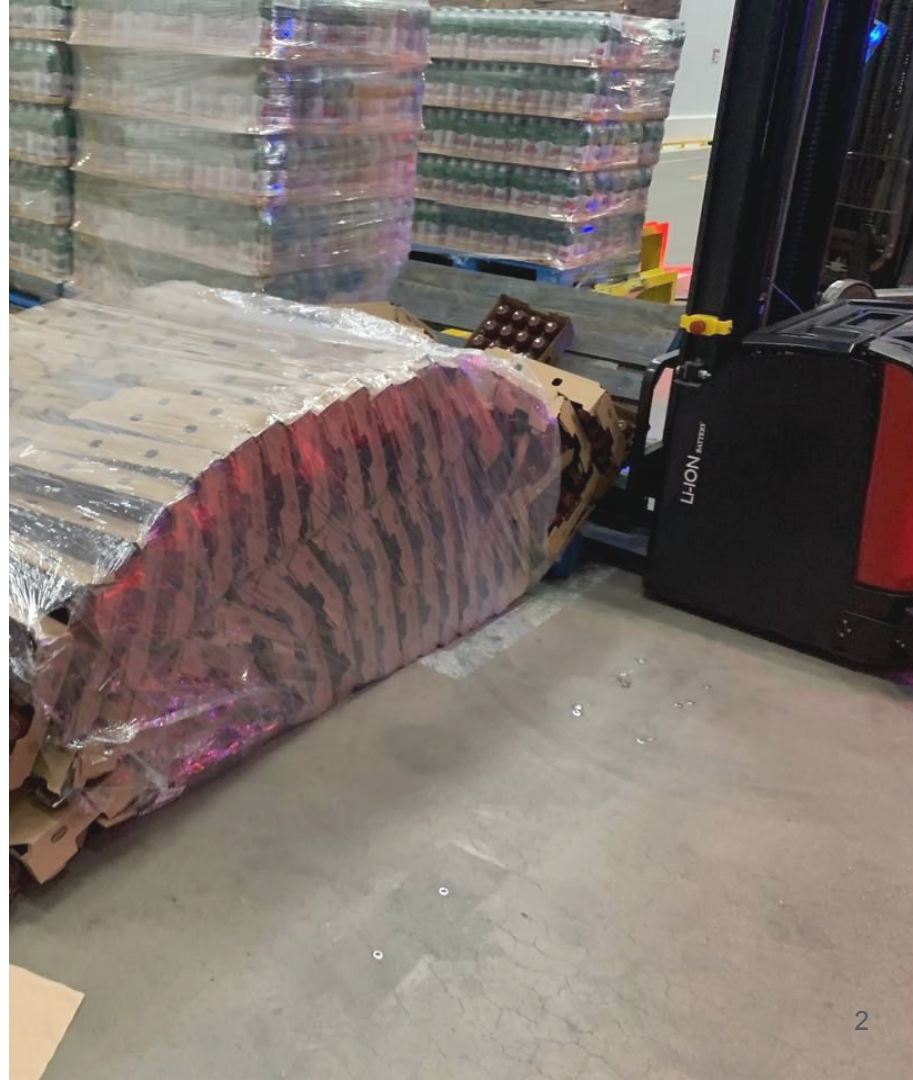
Author: Gabriel F P Araujo

Supervisor: Profa. Genaina Rodrigues

Ricardo Caldas and prof. Claudio Menghi and prof. Patrizio Pelliccione

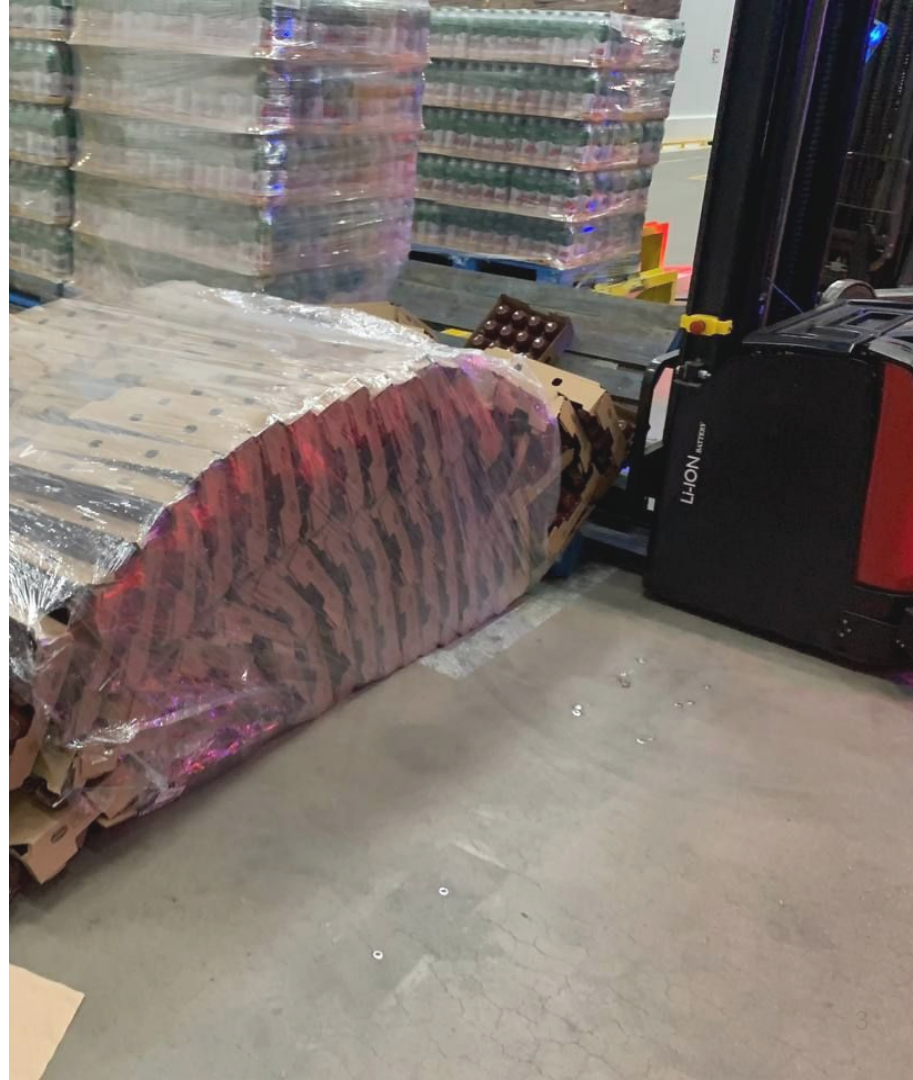
Story time!

- Autonomous forklift fault while picking a pallet.
- We corrected a bug but we forgot to fix one of the parameters



Testing autonomous systems

- Testing robots is expensive
 - Takes time
 - Dangerous
 - Lots of modules to debug
- We need a smart way to test and find problems:
 - Unit tests
 - Integration tests
 - Hardware testing
 - Hardware in the loop



So...

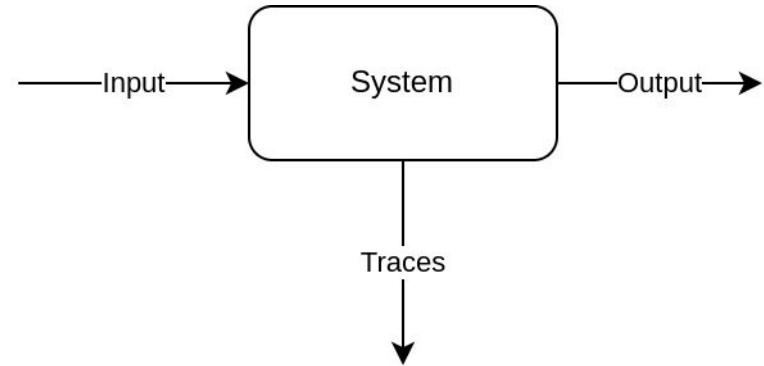
- How do we know that the robots are working?
- How do we know that any change in the codebase will not break other features?
- Are the parameters right?
- Does the different modules work together?

“In the beginning...”

As engineers, we can always:

- **Test** system
- **Record** data
- **Analyse** data

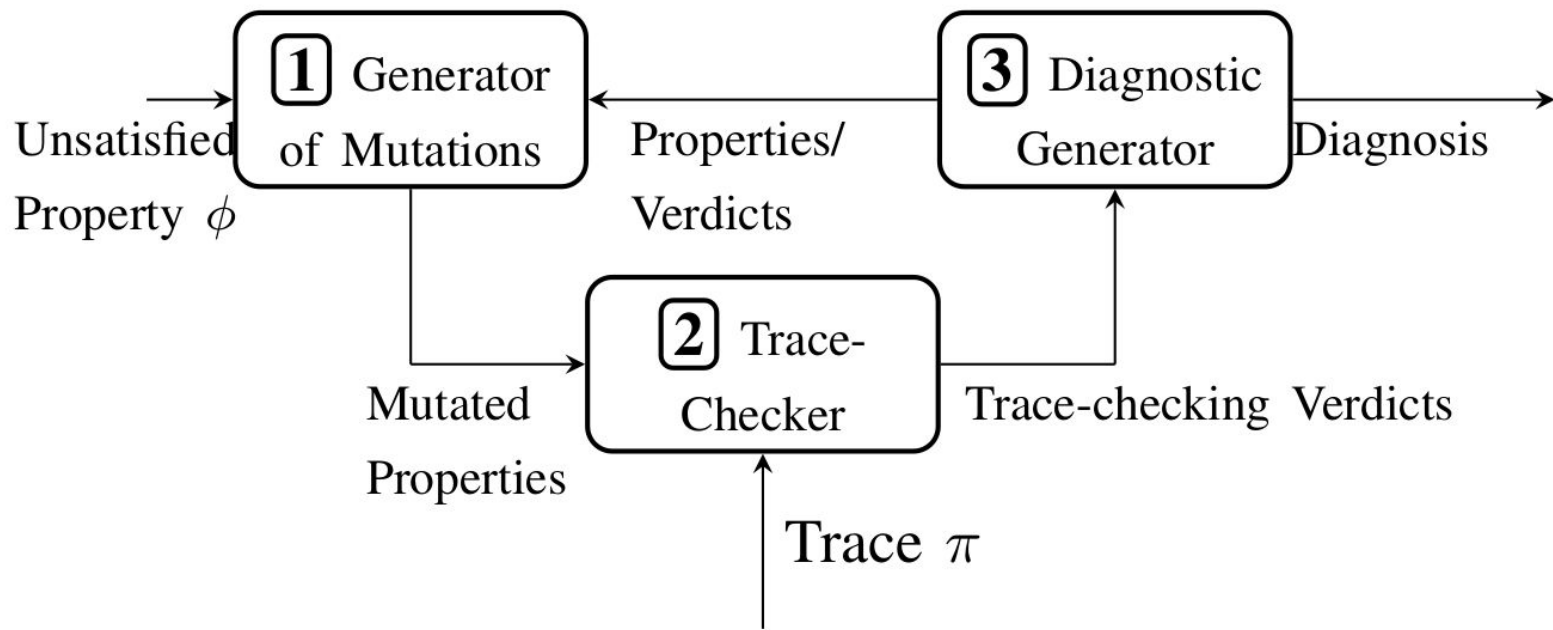
To diagnose *Any* system



Trace-checking

- Engineers record and analyse a system traces to check whether they obey the system's requirements
- We can automate the checking using a tool, a trace-checking tool
- For each property the trace-checking tool outputs a verdict (property satisfied or unsatisfied)

Our Approach



To specify properties: HLS (Hybrid Logic of Signals)

- Extends existing specification languages
 - Time-based languages (e.g. STL) and
 - Sequence-based languages (e.g. LTL)
- Target Cyber-Physical Systems
- Design goals
 - Timestamp variables
 - Index variables
 - Real-valued variables
- ThEodorE is the trace-checker for HLS properties

[1] Menghi, C., Vigano, E., Bianculli, D., & Briand, L. C. (2021). Trace-checking CPS properties: Bridging the cyber-physical gap. *Proceedings - International Conference on Software Engineering*, 847–859. <https://doi.org/10.1109/ICSE43902.2021.00082>

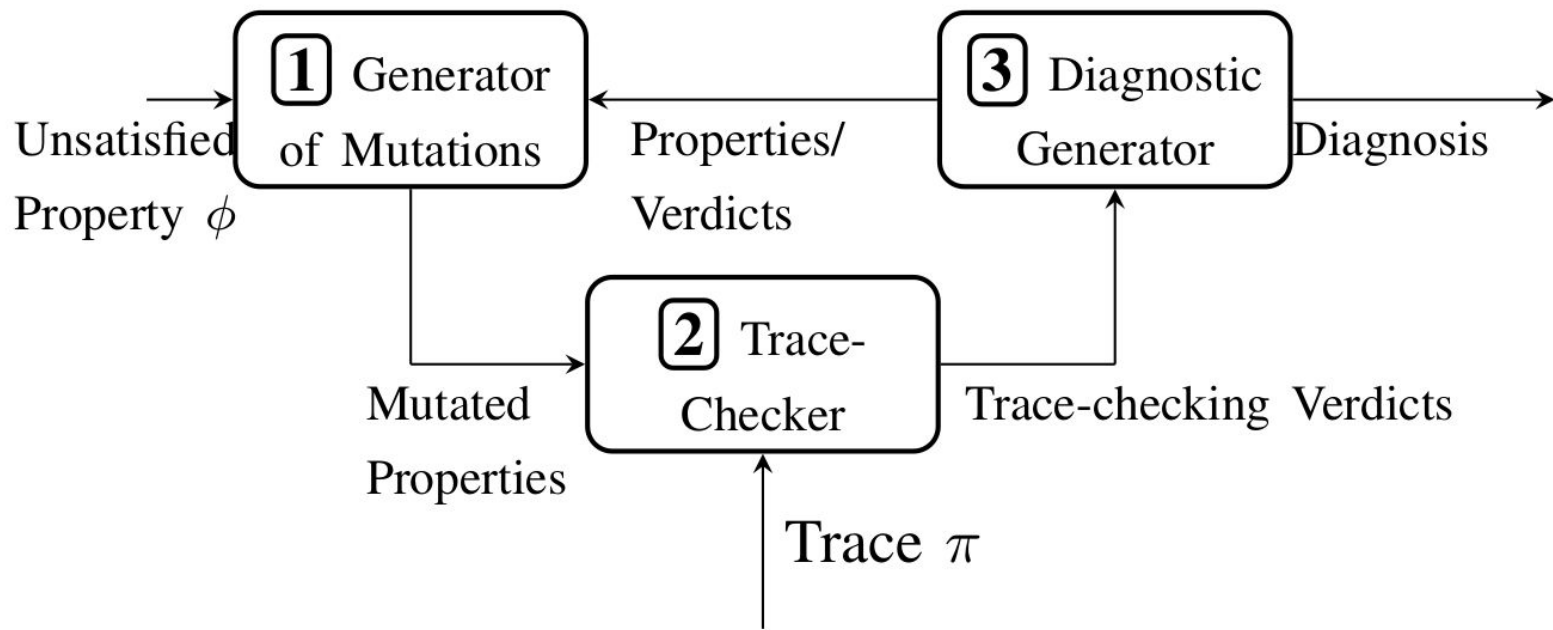
Genetic programming

- Search algorithm based on natural selection
- The genetic algorithm repeatedly modifies a population of individuals
- Used to expand the search space

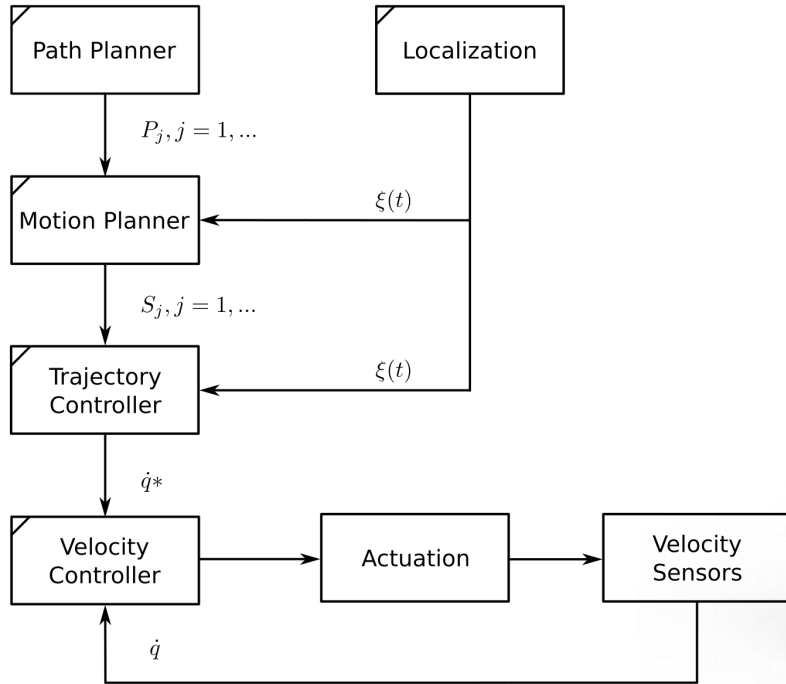
Decision Trees

- Supervised learning algorithm
- Used to classify GA's mutated properties
- To find the root cause in the property

Our Approach



Running example: Autonomous car

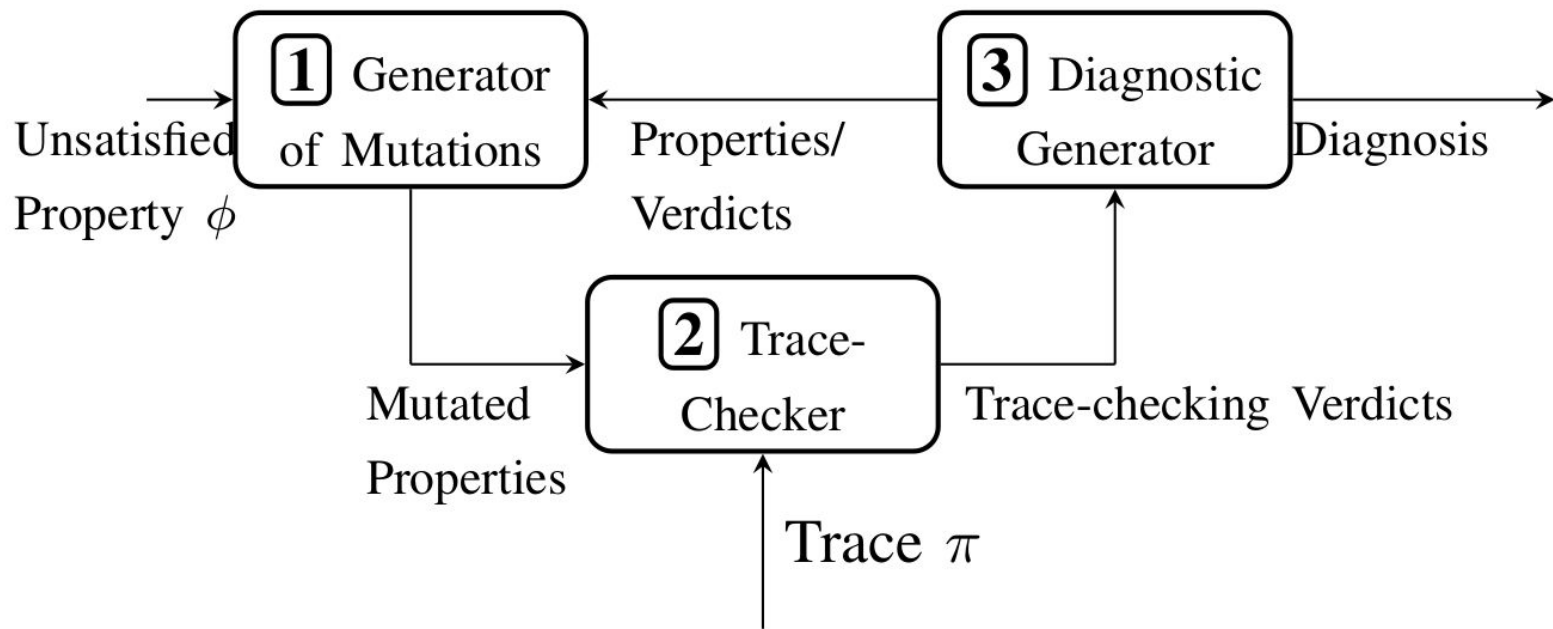


Running example: Requirement

“The car has to follow the desired position in x axis with 20 cm tolerance and its distance to an obstacle must be greater than 45 cm”

$\phi_o ::= \mathbf{forall} \tau_0 \mathbf{in} [0, \infty) : (d_pos_x(\tau_0) - v_pos_x(\tau_0)) > 80\text{cm} \mathbf{and} d2obs(\tau_0) > 45\text{cm}$

Our Approach



Generator of mutations

- Based-on genetic programming
- Creates several versions of the formula changing the terms

$\phi_1 ::= \text{forall } \tau_0 \text{ in } [0, \infty) :$
 $(d_pos_x(\tau_0) - v_pos_x(\tau_0)) > 80\text{cm}$
 and $d2obs(\tau_0) > 45\text{cm}$

$\phi_2 ::= \text{forall } \tau_0 \text{ in } [0, \infty) :$
 $(d_pos_x(\tau_0) - v_pos_x(\tau_0)) > 20\text{cm}$
 and $d2obs(\tau_0) < 45\text{cm}$

Trace Checker

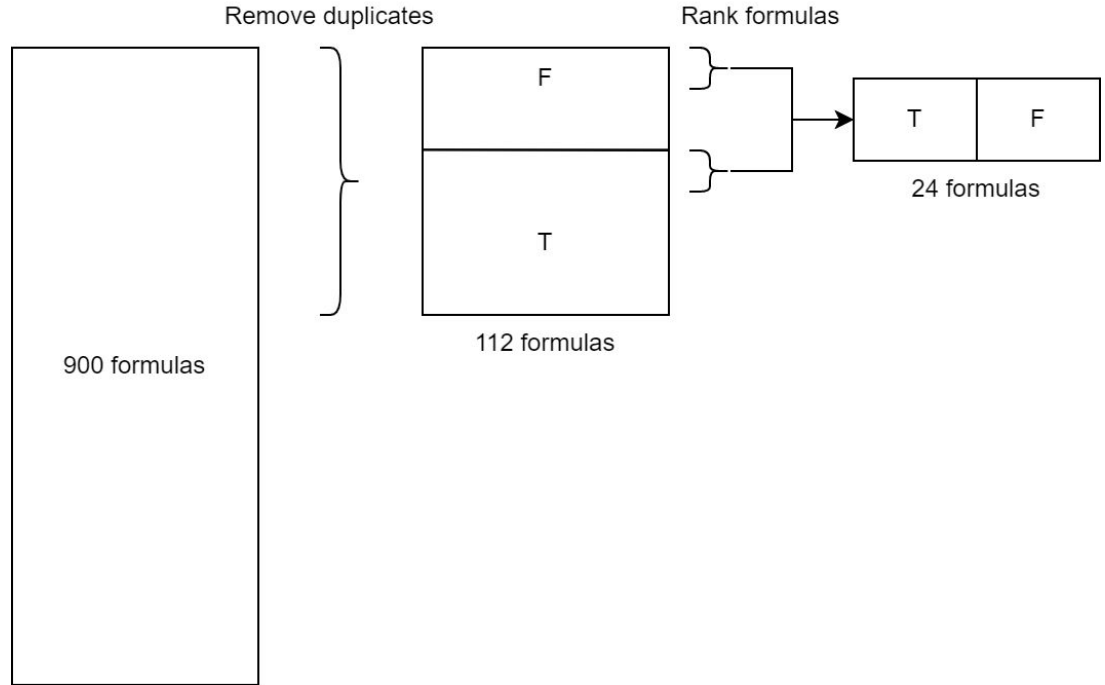
- For each generated formula, we check them using ThEodorE

$\phi_1 ::= \text{forall } \tau_0 \text{ in } [0, \infty) :$
 $(d_pos_x(\tau_0) - v_pos_x(\tau_0)) > 80\text{cm} \longrightarrow \text{False}$
 and $d2obs(\tau_0) > 45\text{cm}$

$\phi_2 ::= \text{forall } \tau_0 \text{ in } [0, \infty) :$
 $(d_pos_x(\tau_0) - v_pos_x(\tau_0)) > 20\text{cm} \longrightarrow \text{True}$
 and $d2obs(\tau_0) < 45\text{cm}$

Diagnostic Generator

After, we run a lot of new formulas (>1000), we choose the best ones based on how distant they are from the **original** formula.



DG: Ranking mutated properties

We use the Smith-Waterman algorithm to calculate the similarity between two formulas.

Formulas

Score

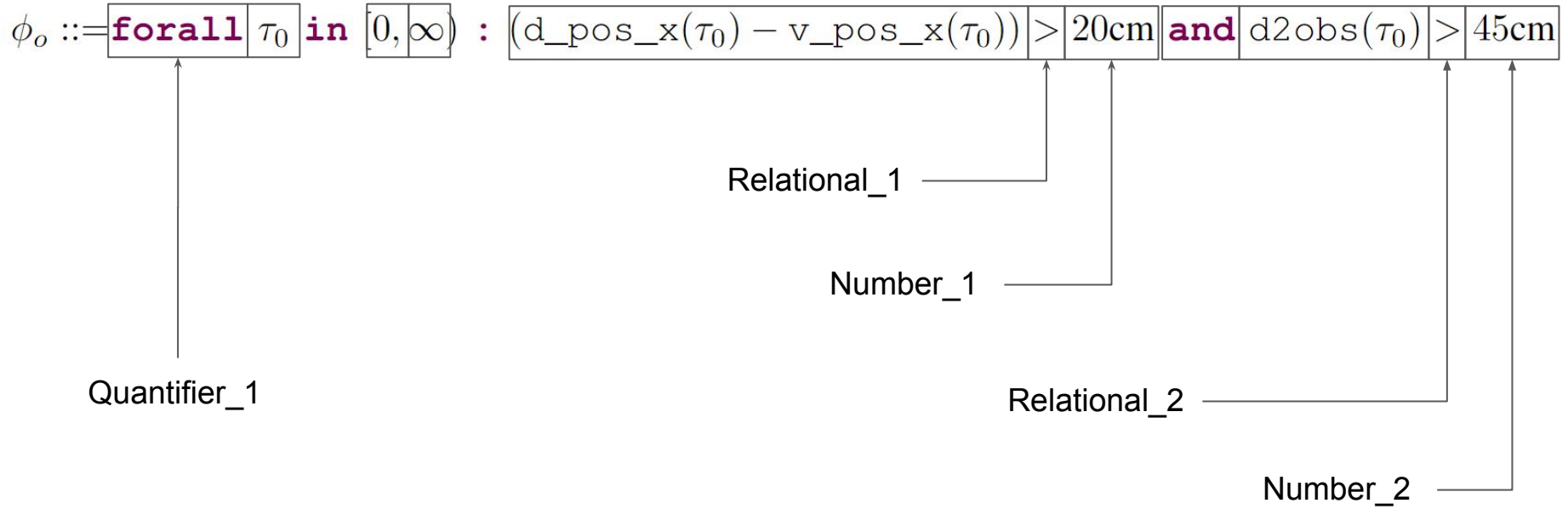
$\phi_o ::= \mathbf{forall} \tau_0 \mathbf{in} [0, \infty) : (d_pos_x(\tau_0) - v_pos_x(\tau_0)) > 20\text{cm} \mathbf{and} d2\text{obs}(\tau_0) > 45\text{cm}$	–
$\phi_1 ::= \mathbf{forall} \tau_0 \mathbf{in} [0, \infty) : (d_pos_x(\tau_0) - v_pos_x(\tau_0)) > 80\text{cm} \mathbf{and} d2\text{obs}(\tau_0) > 45\text{cm}$	25
$\phi_2 ::= \mathbf{forall} \tau_0 \mathbf{in} [0, \infty) : (d_pos_x(\tau_0) - v_pos_x(\tau_0)) > 20\text{cm} \mathbf{and} d2\text{obs}(\tau_0) < 45\text{cm}$	25
$\phi_2 ::= \mathbf{forall} \tau_0 \mathbf{in} [0, \infty) : (d_pos_x(\tau_0) - v_pos_x(\tau_0)) > 20\text{cm} \mathbf{or} d2\text{obs}(\tau_0) < 45\text{cm}$	20

MATCH = 3

MISMATCH = -3

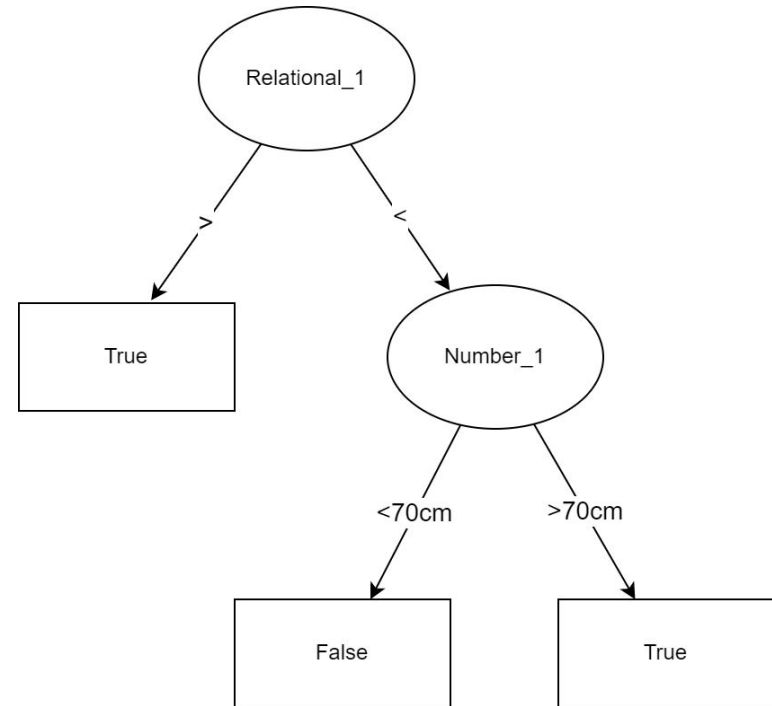
Diagnostic Generator

We label the terms in the formula:



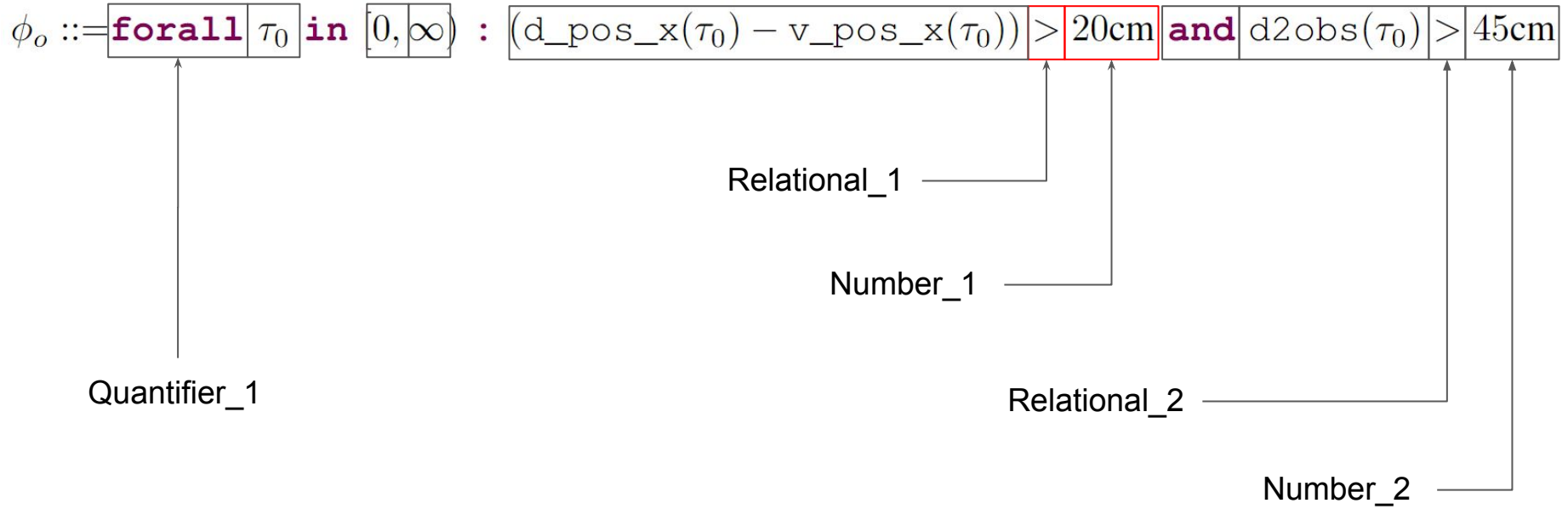
Diagnostic Generator

- Use J48 from WEKA to learn which terms change when verdict also changes
 - Unsupervised learning
 - Output a decision tree
 - Extract which terms on the formula are most relevant

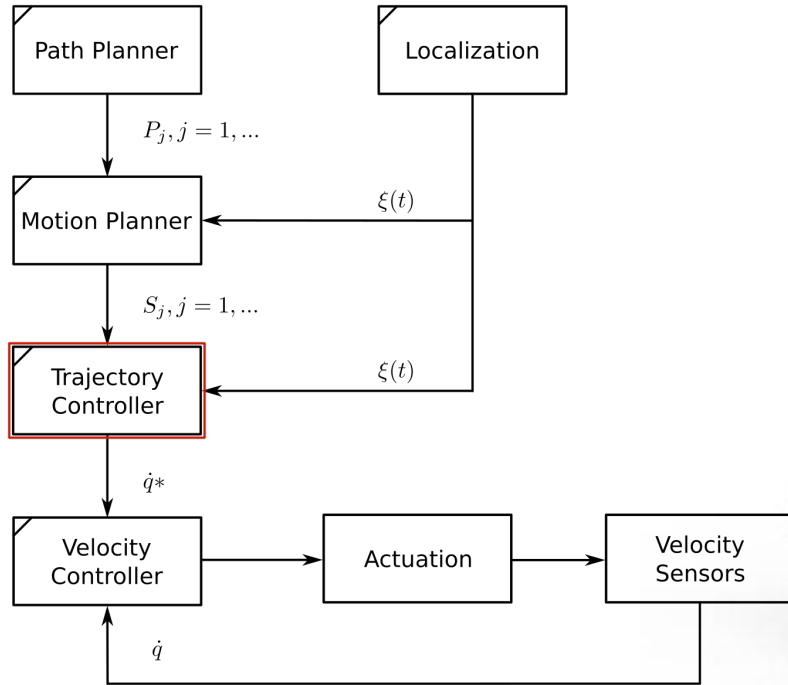


Diagnostic Generator

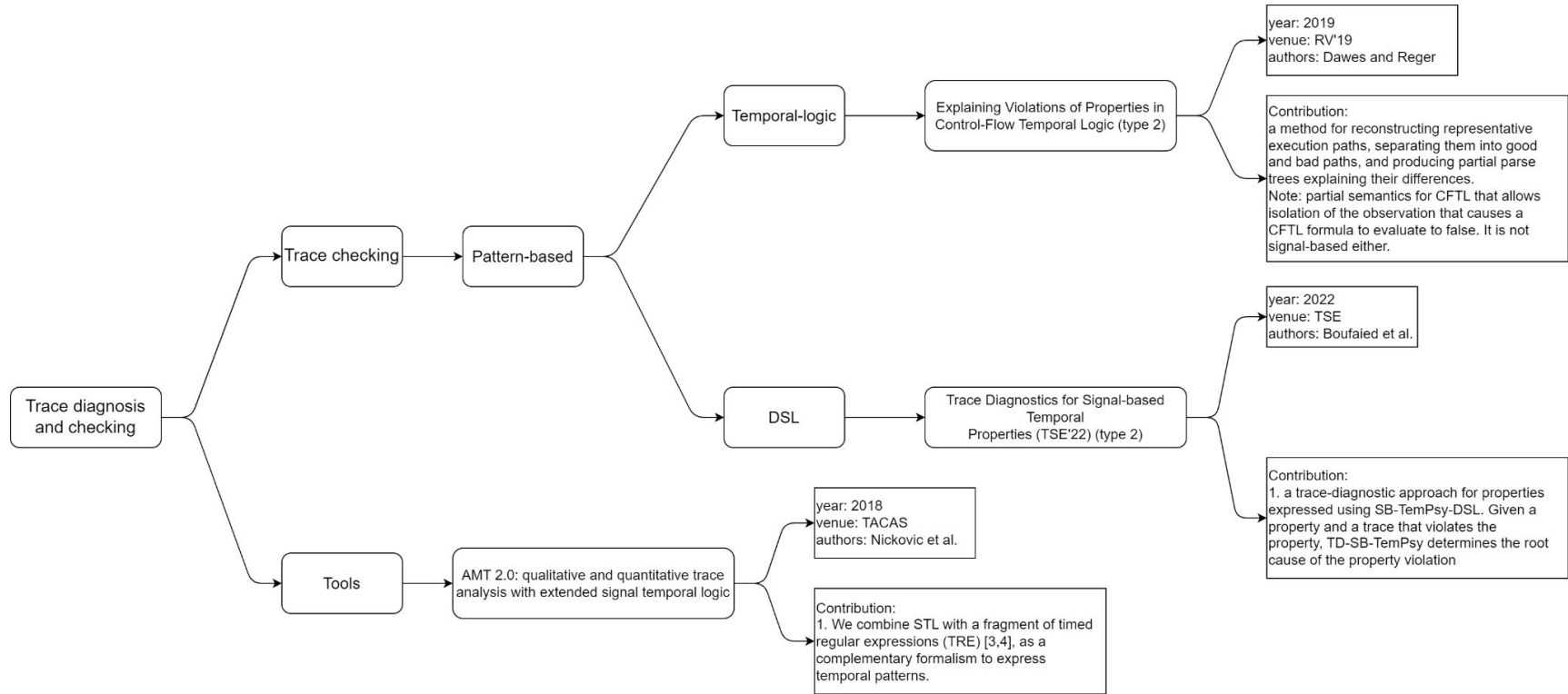
We label the terms in the formula:



Running example



Related work



Conclusions and what is next?

- We achieved so far:
 - Definition of the problem
 - Running example
 - Approach working
- Running the experiments for the evaluation.

Thank you