Tree based diagnosis enhanced with meta knowledge *Diagnosis indicator discovery as a by product* 

*Louis Goupil, Elodie Chanthery, Louise Travé-Massuyès, Sébastien Delautier (ATOS)* 





# Goal: Diagnosis and correction of a system without model

**Diagnosis** of a system without physical model can be achieved with **Data-Based** methods

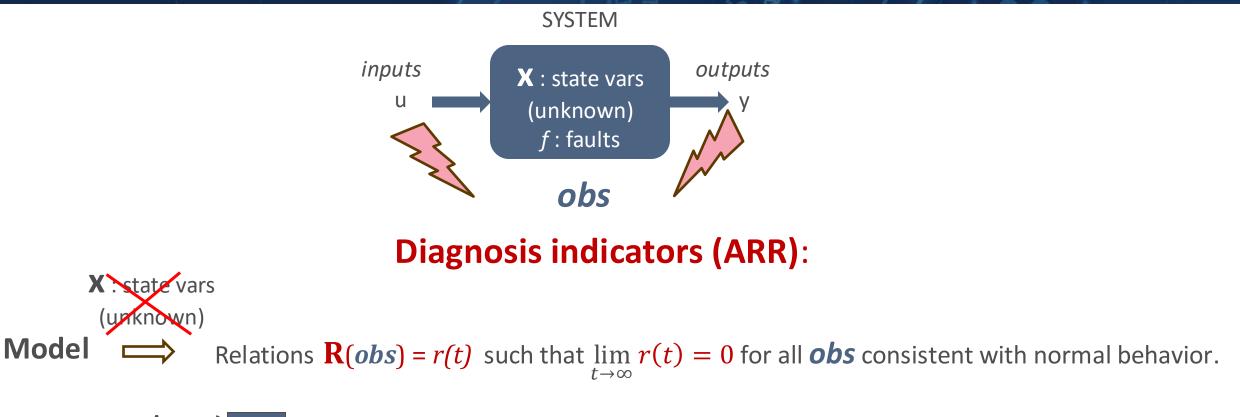
**Correction** of a faulty system requires **isolation/explanation** of the fault

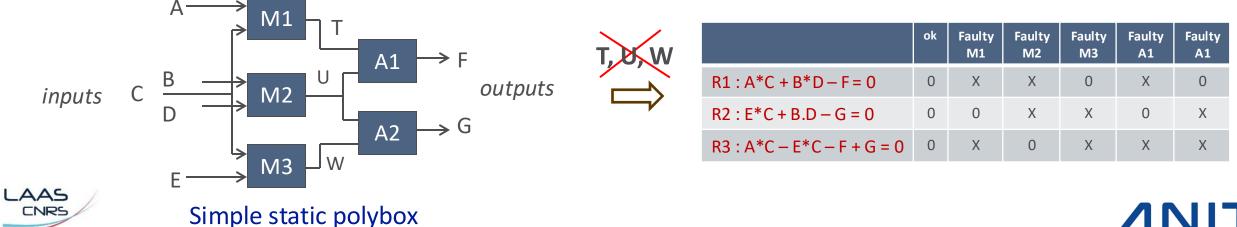
→ Data provides meaninfull information but important knowledge can be leveraged from Model-Based Diagnosis



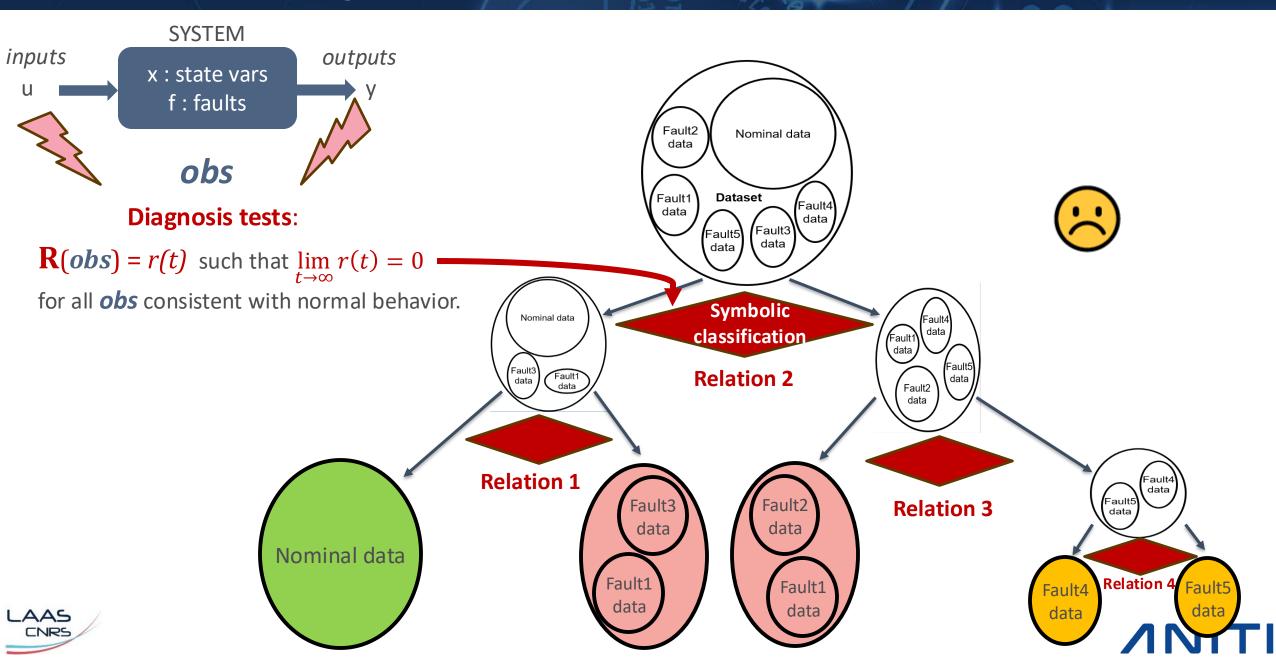


#### Model based diagnosis knowledge

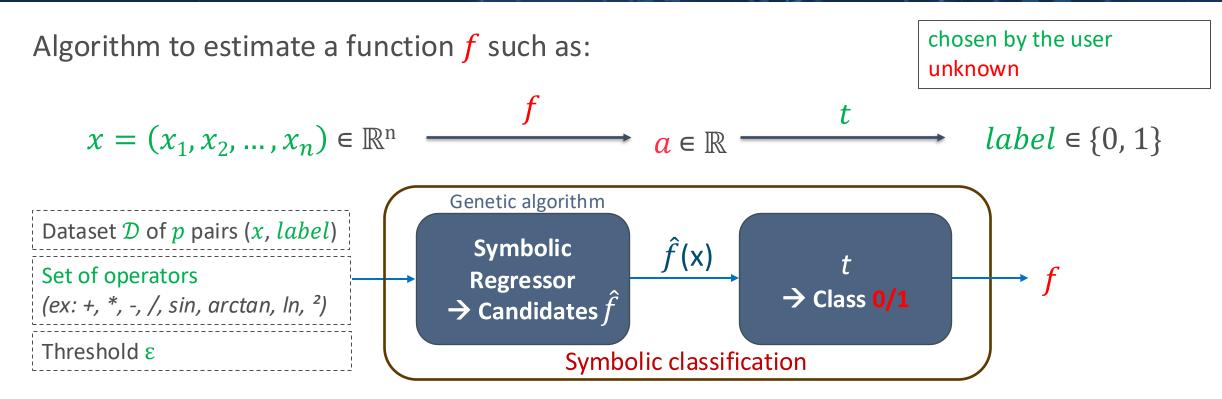




#### Data based diagnosis: multivariate decision trees



#### Symbolic classification



Evaluation of the candidate c with the log-loss fitness function

$$Fitness(\hat{f}) = -\frac{1}{p} \sum_{(x,label) \in \mathcal{D}} [label * \ln(t(c(x))) + (1 - label) * \ln(1 - t(c(x)))]$$

#### **Stopping condition**

IF  $Fitness(\hat{f}) < \varepsilon$  THEN algorithm stops and f is « found » ELSE  $Mutate(\hat{f})$  AND Repeat



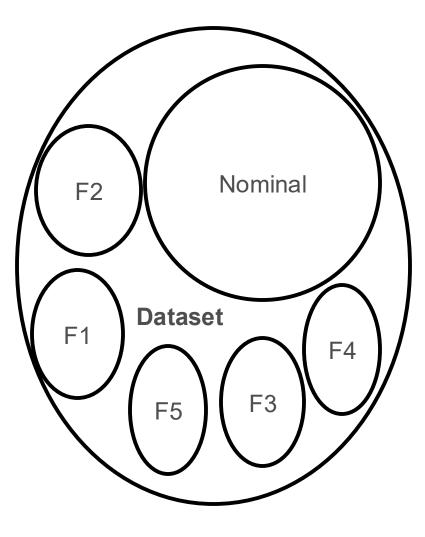
## **DT4X: Diagnosis Tree for eXplainability**



Train a symbolic classifier to find f



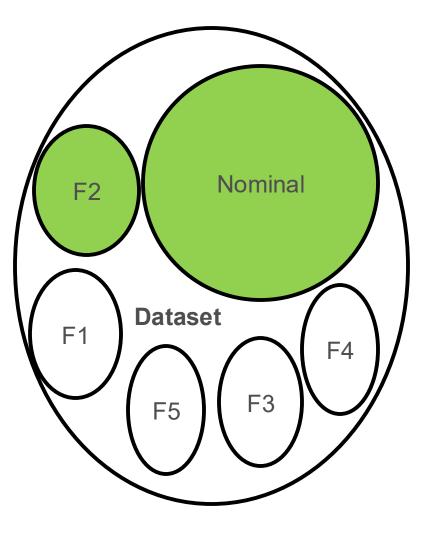






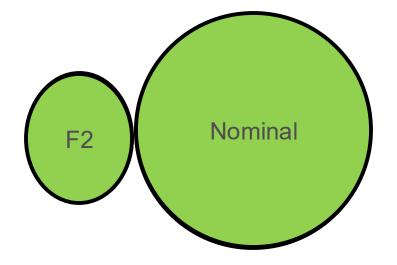


Select 2 classes (preferably one being the nominal class)



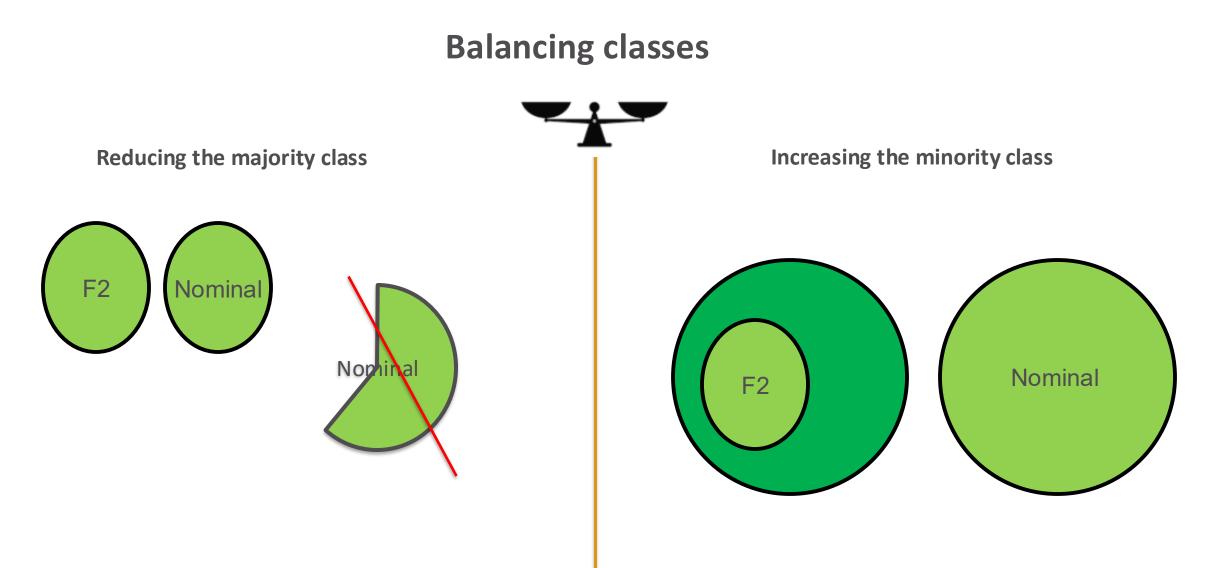






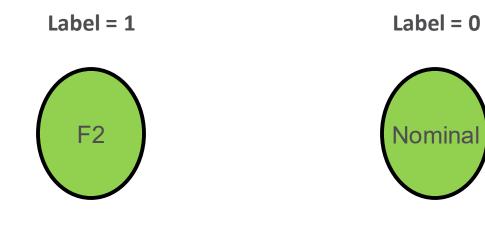








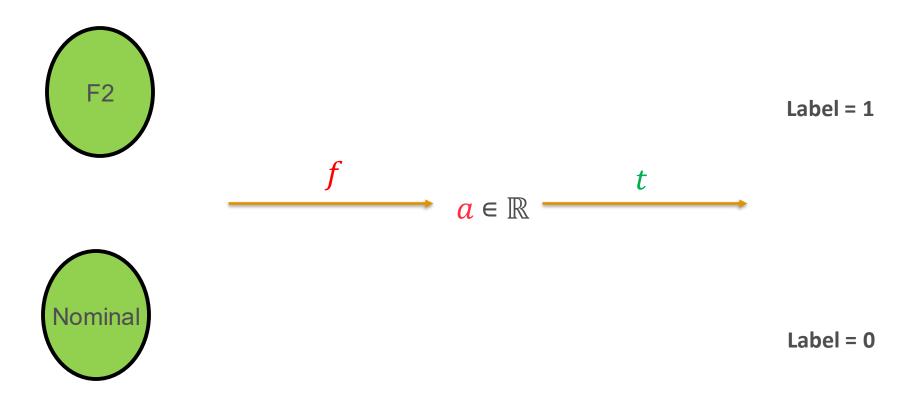






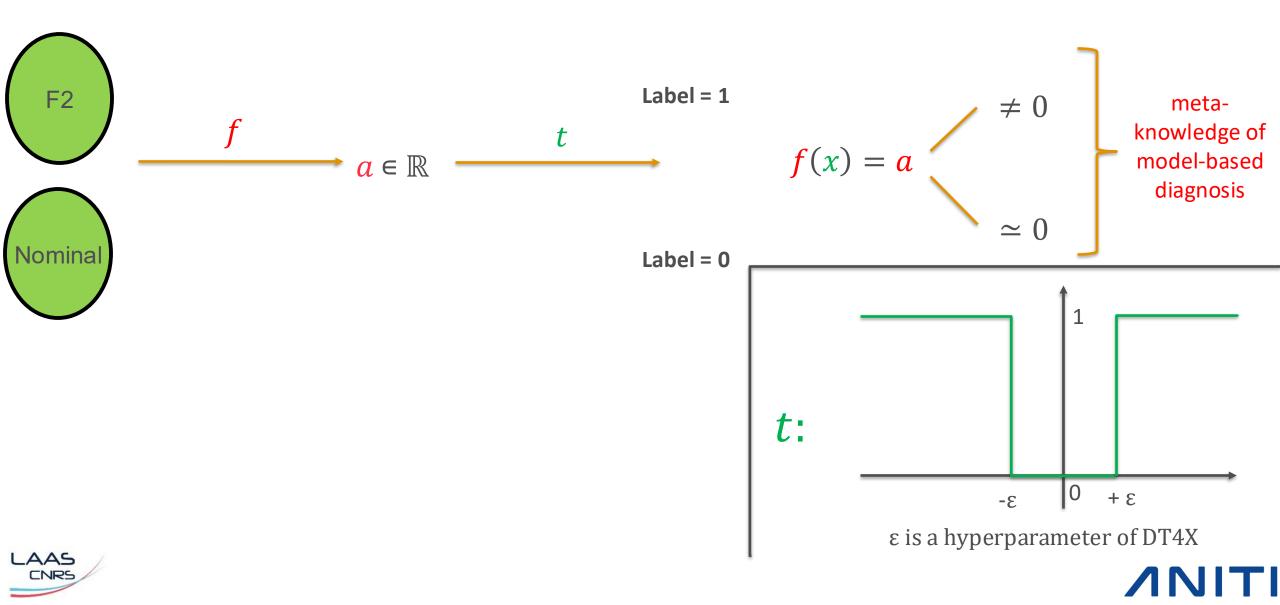


Use **symbolic classification** to find *f* 





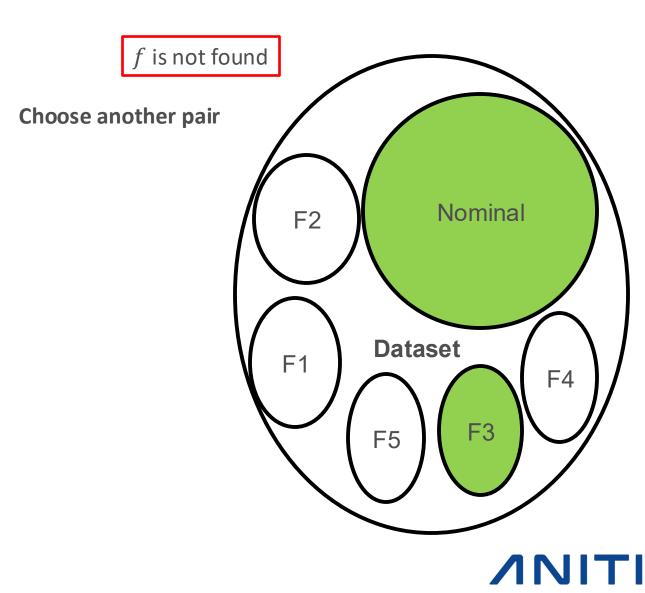




f is not found



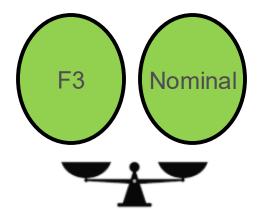






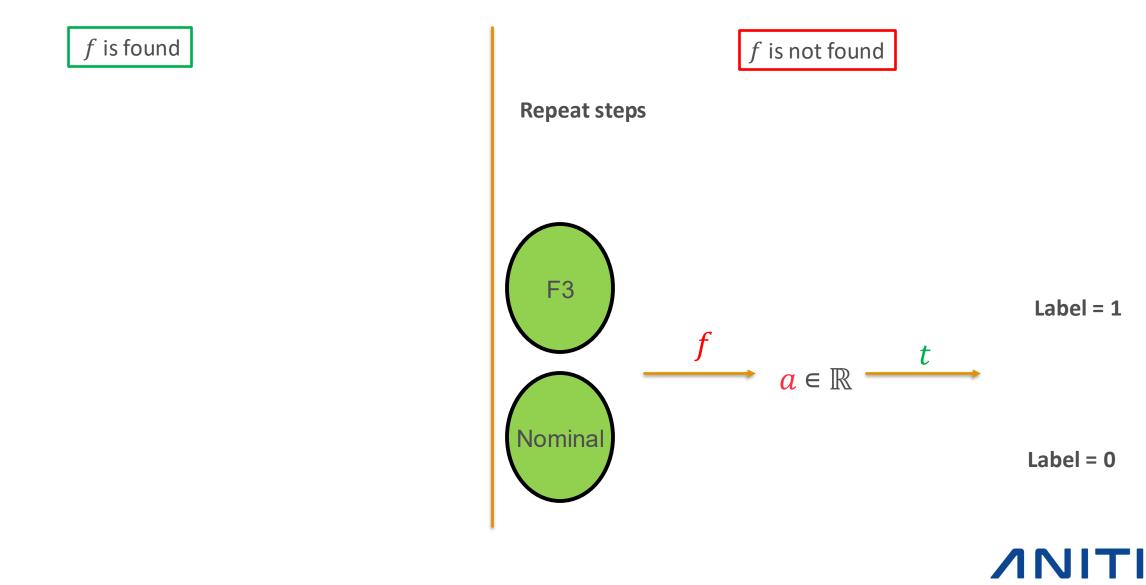


**Repeat steps** 

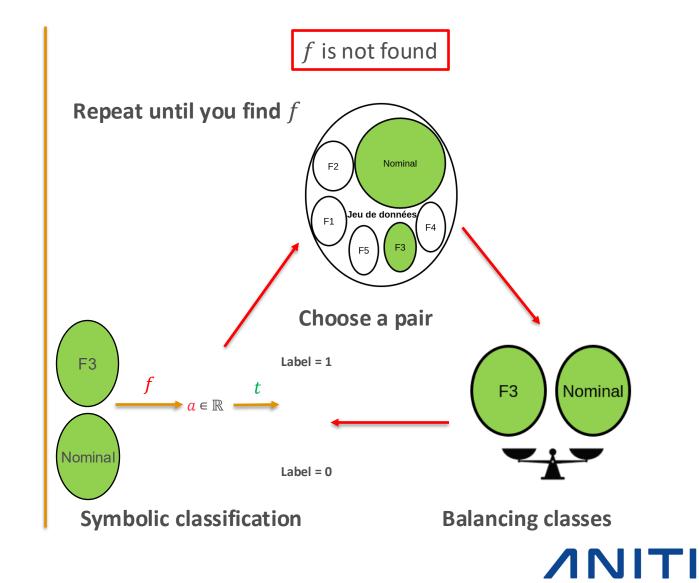














f is not found

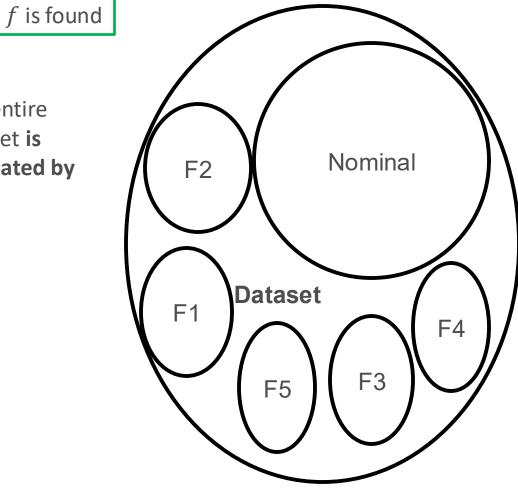
IF all pairs have been tried on AND f is not found

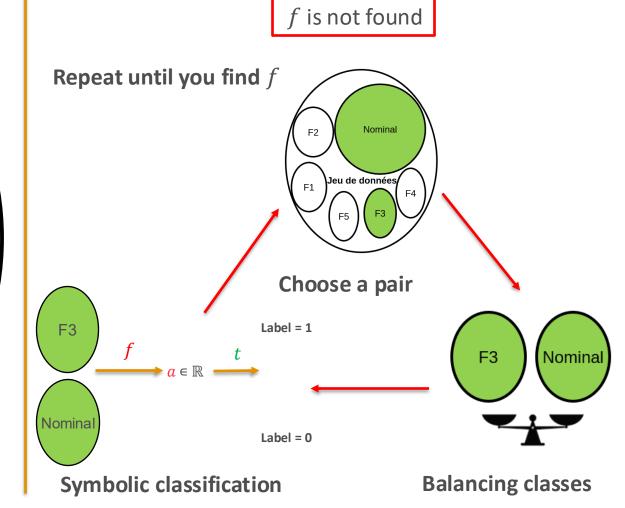
> THEN Stop the algorithm





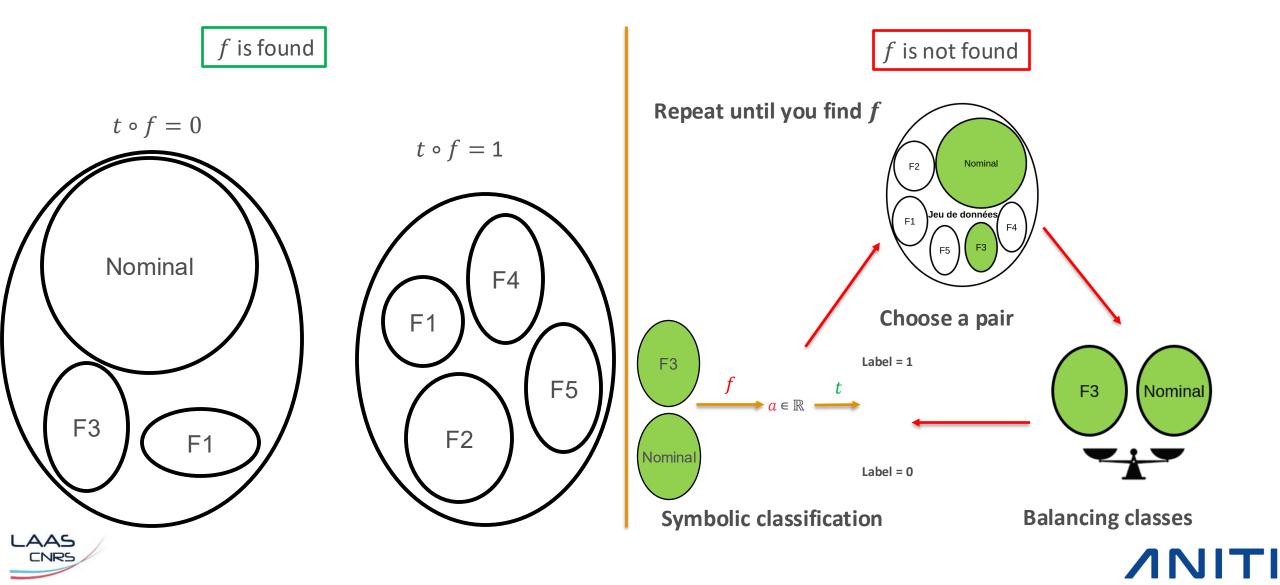
The entire dataset is evaluated by  $t \circ f$ 

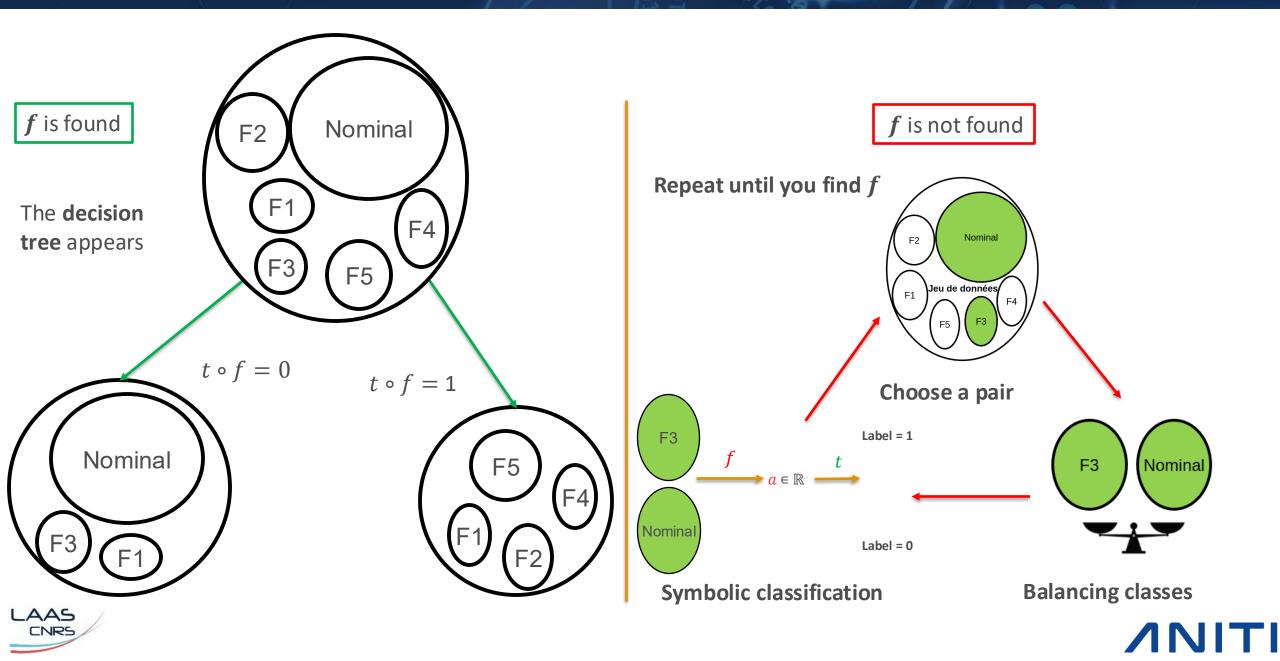


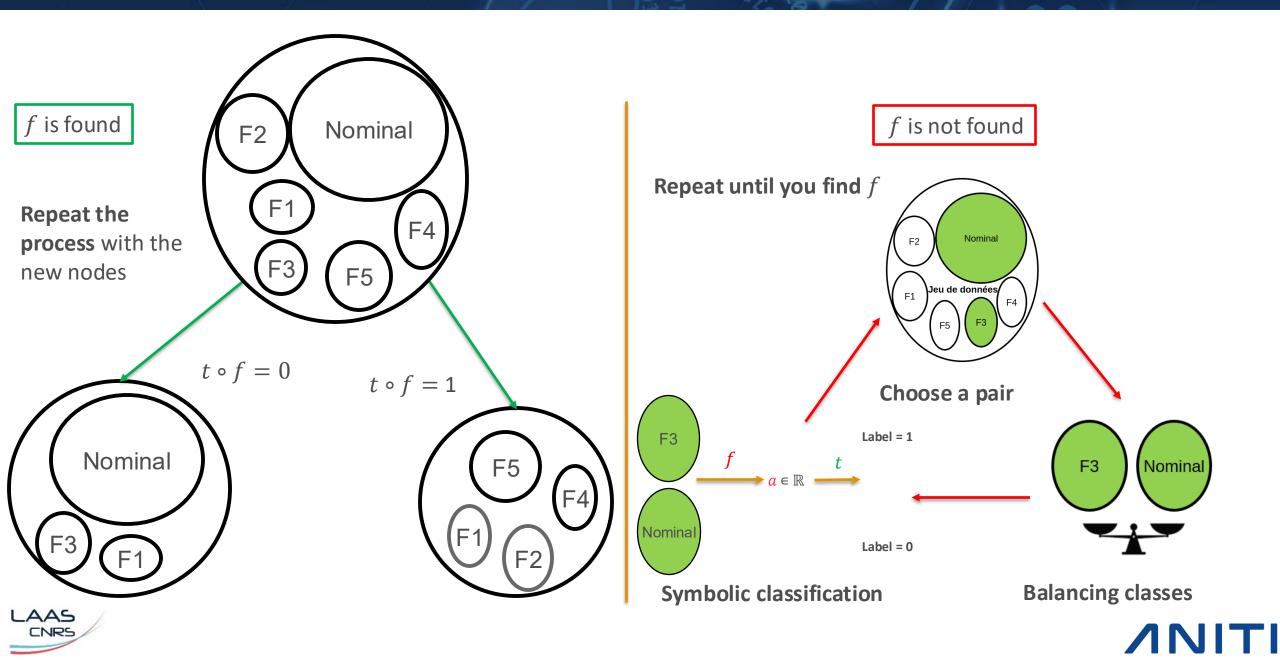


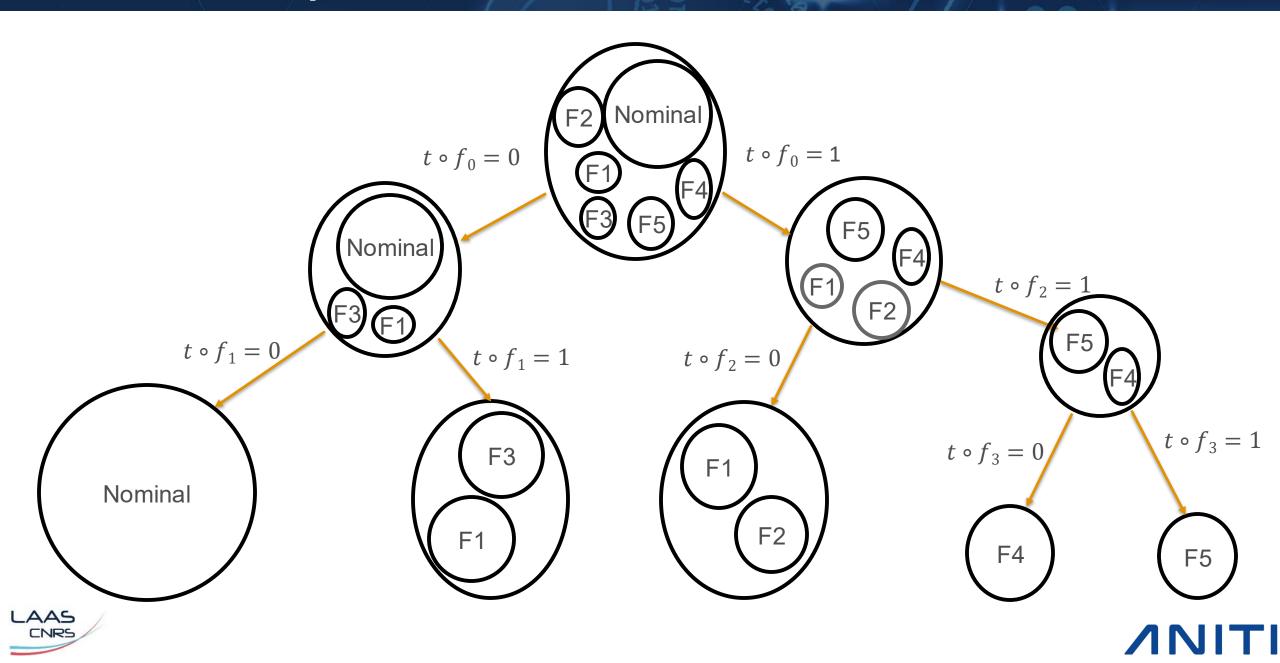
**NNITI** 



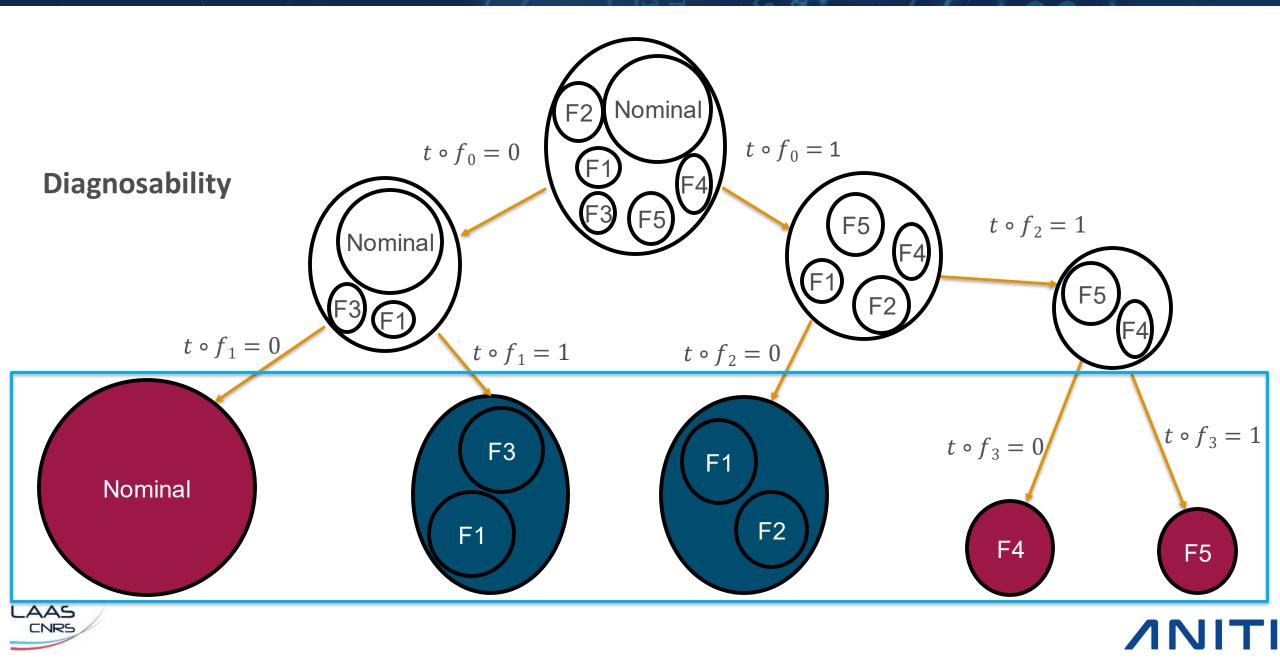


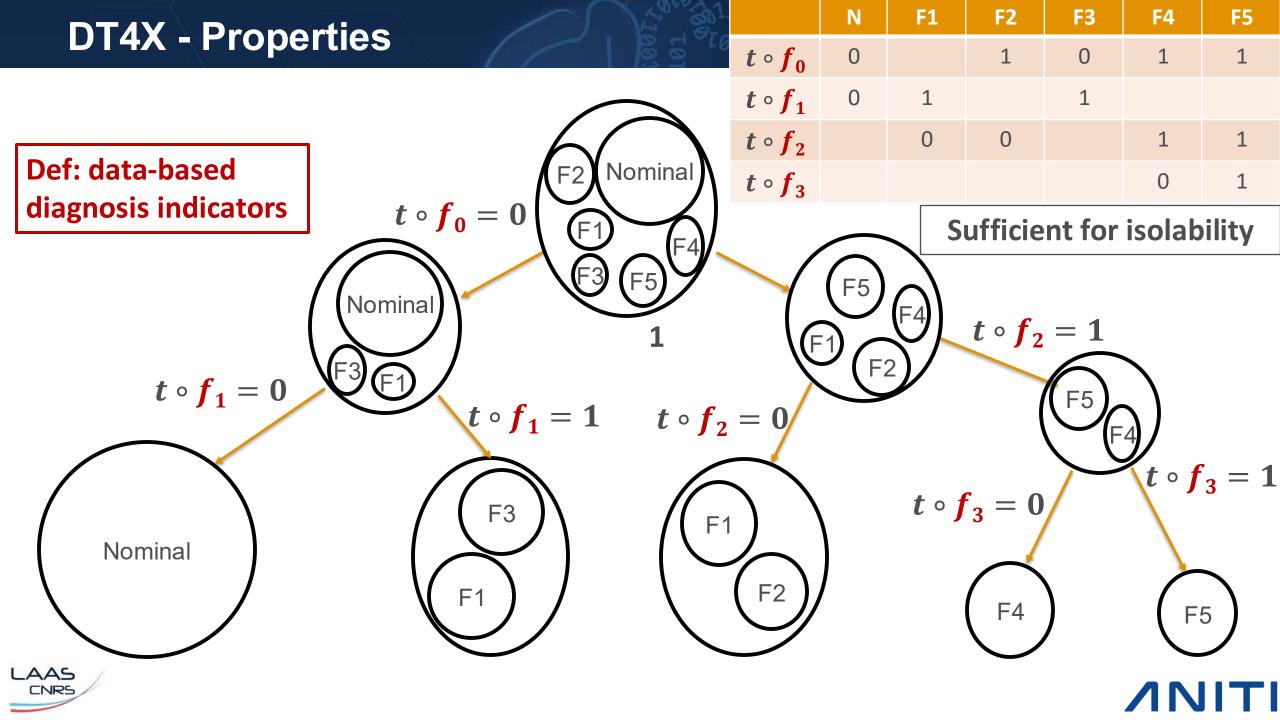




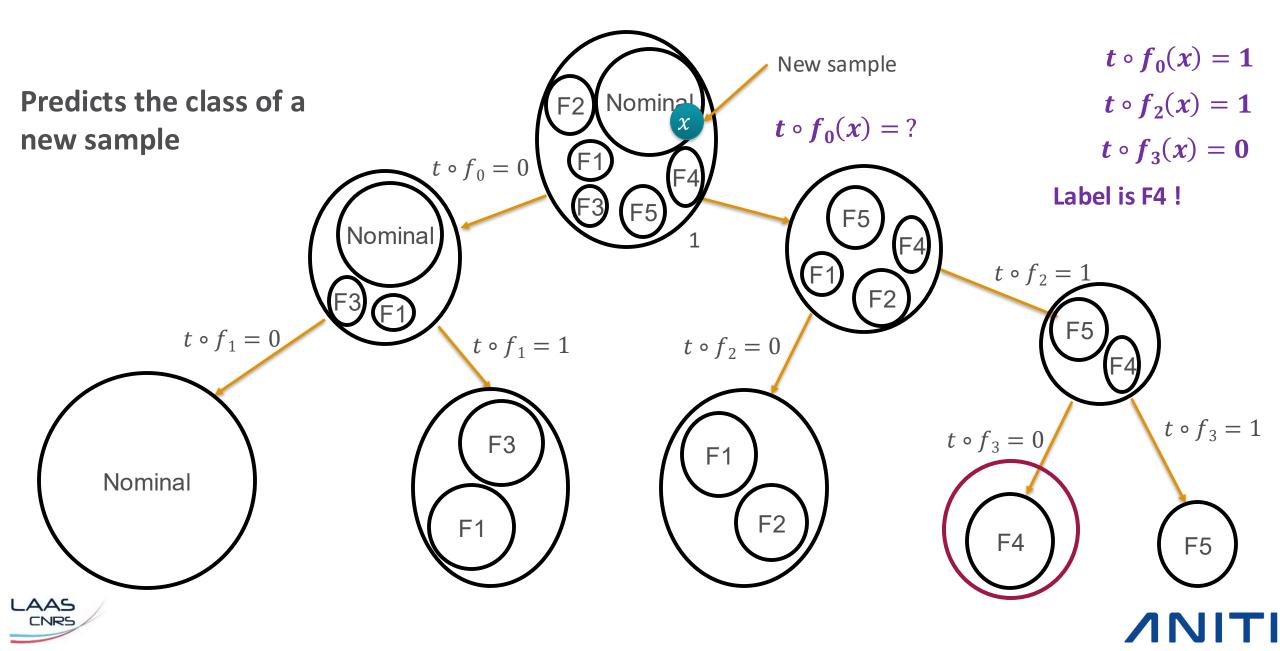


#### **DT4X - Properties**





### **DT4X - Properties**



Two data-based ARRs on the same path in the tree generated by DT4X are different:

 $\boldsymbol{f}_1 \neq \boldsymbol{f}_2$ 

ARR : **f**<sub>1</sub>  $D_1$ ARR :  $f_2$  $D_2$ 

Root Node

D

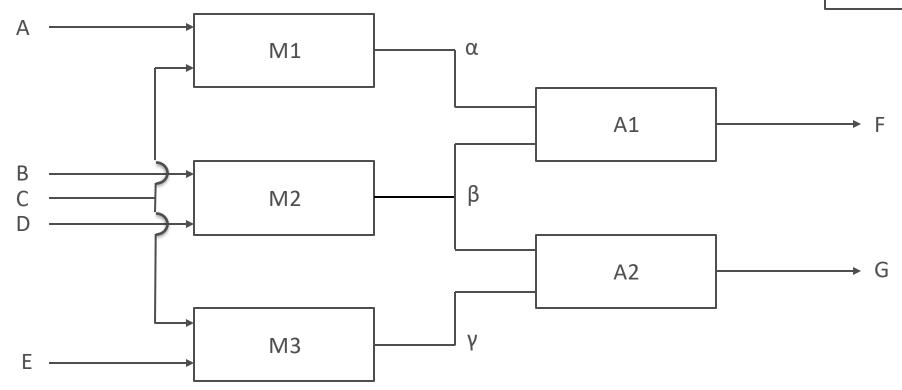
Responds to tree optimality issues, no superfluous nodes





Polybox

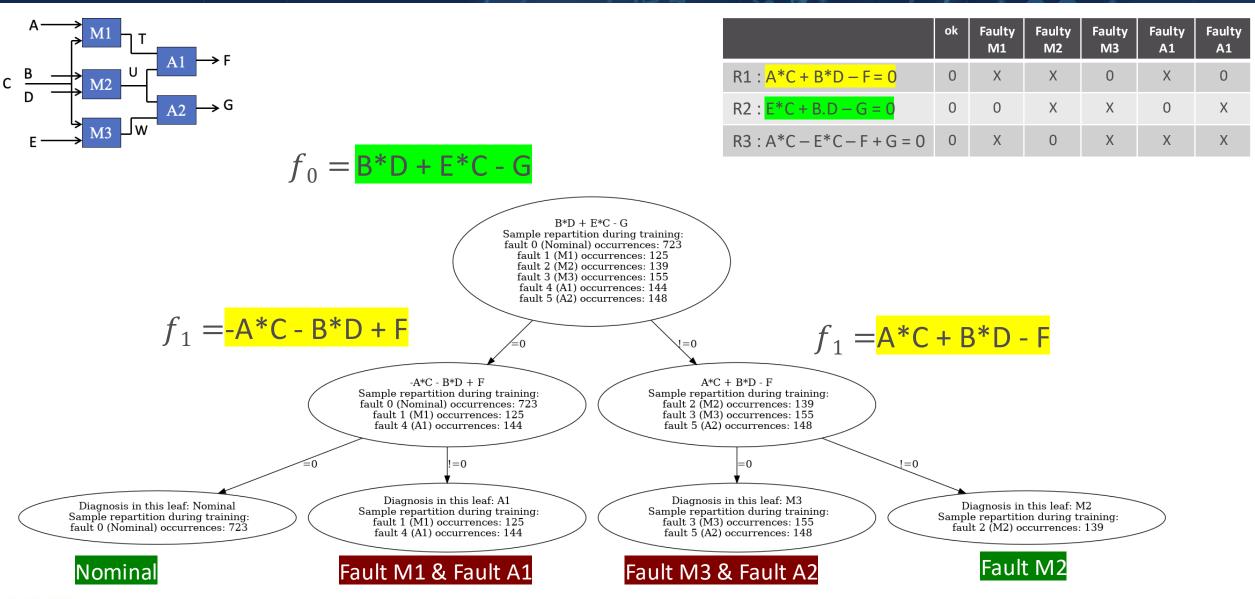
Type: static Known ARRs: yes Data: simulated







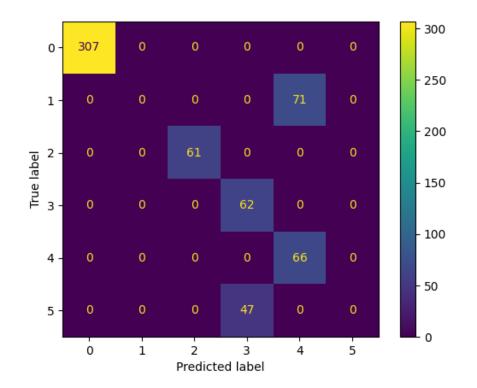
#### **PolyBox Results – Single Faults**



LAAS

#### ΛΝΙΤΙ

Algorithm	Scoring Time (s) 614 samples	Accuracy (%)	f1-score $(\%)$
DT4X <b>*</b>	0.04	80.78	74.25
sklDT	0.00	46.91	46.02
sklRF	0.01	46.09	38.41
sklLR	0.00	50.00	33.33
sklNB	0.00	49.67	34.17
sklSVM	0.02	50.16	33.69
sklKNN	0.01	48.05	37.43

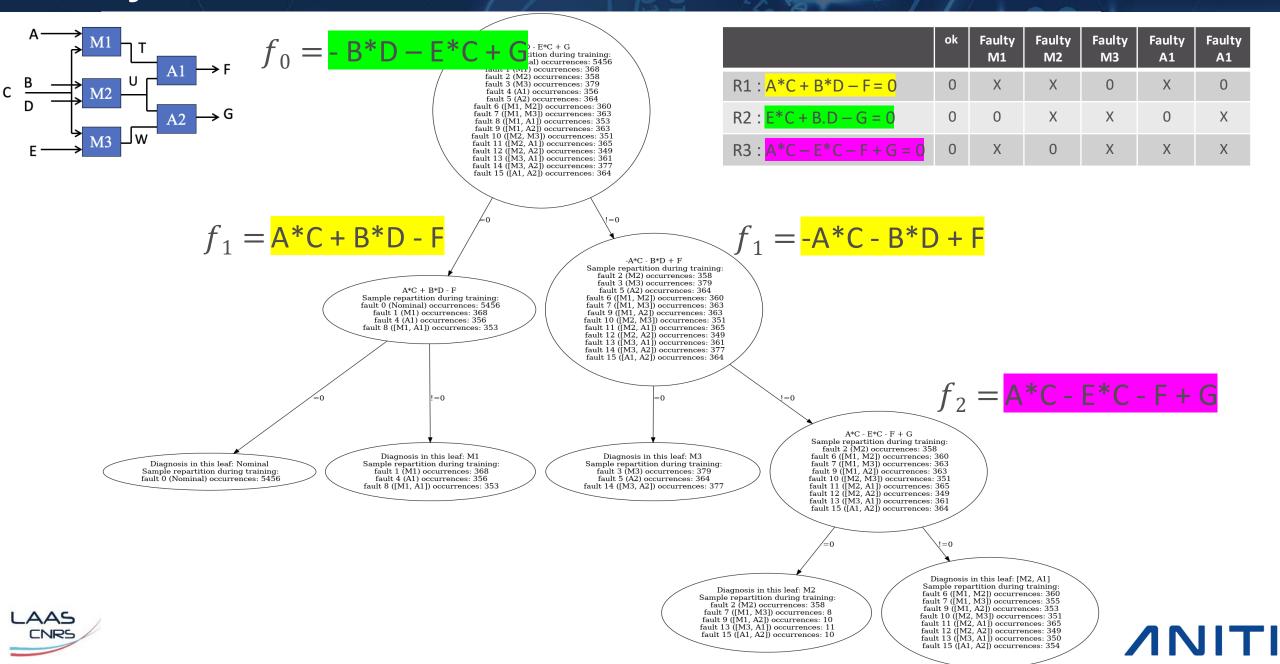


\* DT4X also finds the analytical expression of model based RRAs

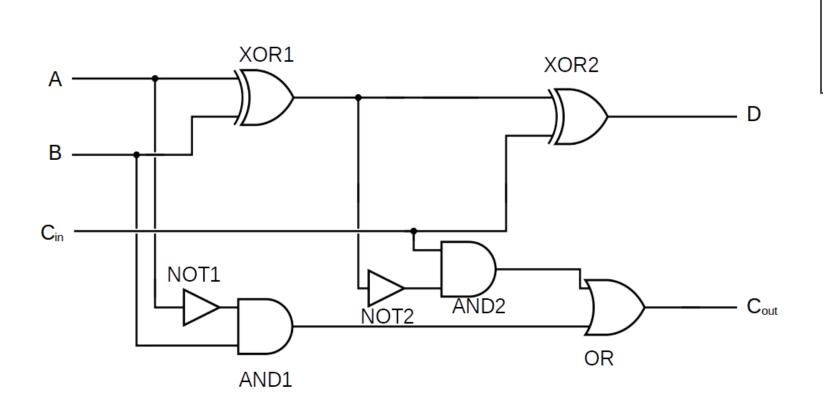




#### **PolyBox Results – Double Faults**



#### **Binary subtractor**



Type: static Known ARRs: no Data: simulated

Fault: output stuck at 0 or 1 Faulty samples behave like nominals

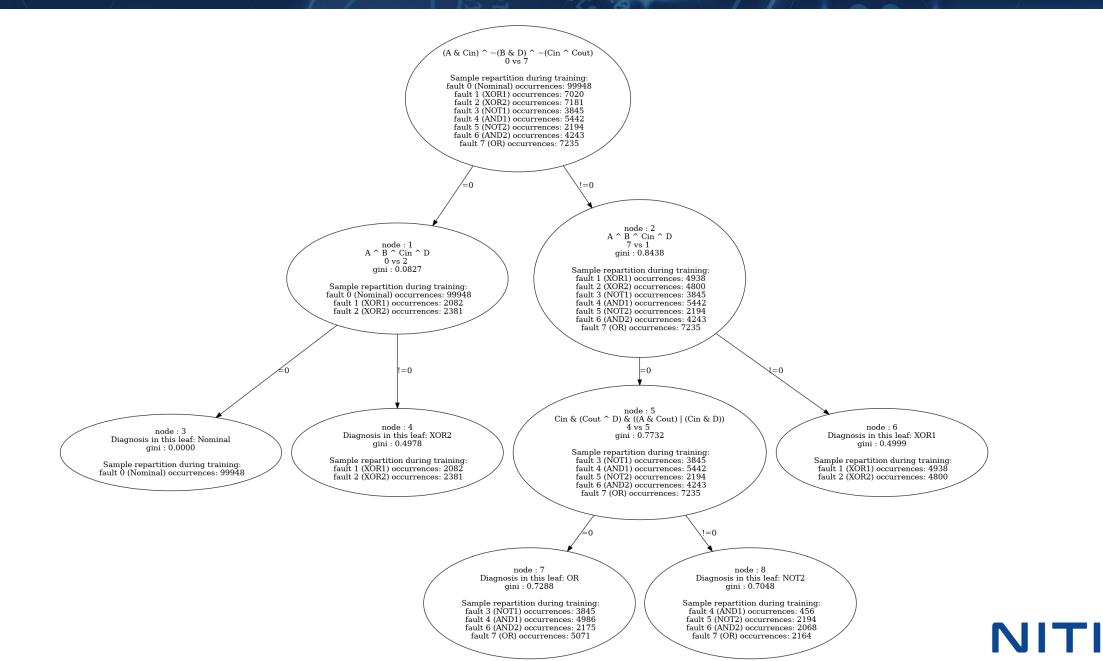
Solution: data pre-processing (from 125,000 to 46384 faulty samples)





#### **Binary subtractor**

\_AAS



	Inpu		Outputs		Diagnosis Indicators				
A	В	$C_{in}$	D	$C_{out}$	$(A\&C_{in})$ $\neg (B\&D)$ $\neg (C_{in}\&C_{out})$	$ \begin{array}{c} A^{}B\\ {}^{}C_{in}{}^{}D\end{array} $	$ \begin{array}{c} C_{in} \& (C_{out} D) \& \\ ((A\&C_{out}) \mid (C_{in}\&D)) \end{array} $		
0	0	0	0	0	0	0	0		
0	0	1	1	1	0	0	0		
0	1	0	1	1	0	0	0		
0	1	1	0	1	0	0	0		
1	0	0	1	0	0	0	0		
1	0	1	0	0	0	0	0		
1	1	0	0	0	0	0	0		
1	1	1	1	1	0	0	0		

: XOR & : AND | : OR

**NITI** 



	Inputs			Out	puts	Expression				
			$f_{XOR1}$ $f_{XOR2}$		KOR2	$f_{XOR1}$	$f_{XOR2}$			
A	B	$C_{in}$	D	$C_{out}$	D	$C_{out}$	$A^B^$	$C_{in}  \hat{D}$		
0	0	0	1	0	1	0	1	1		
0	0	1	0	0	0	1	1	1		
0	1	0	0	1	0	1	1	1		
0	1	1	1	1	1	1	1	1		
1	0	0	0	0	0	0	1	1		
1	0	1	1	1	1	0	1	1		
1	1	0	1	0	1	0	1	1		
1	1	1	0	0	0	1	1	1		

: XOR & : AND | : OR

ΛΝΙ

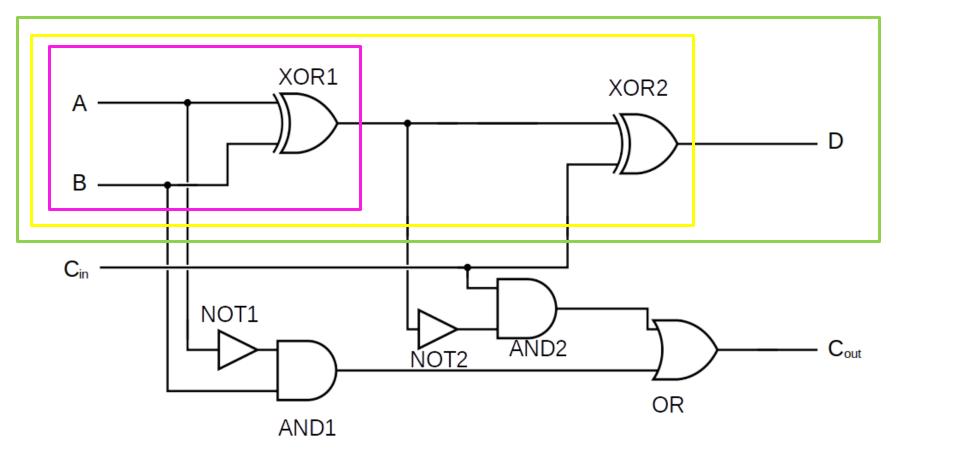
ТΙ

**Def : Logical ARR** := Data-based ARR whose operators are logic gates

#### **Binary subtractor**

 $A^B^Cin^D$ 

A^B^Cin^D A^B^Cin^D A^B^Cin^D



Input-output relationships like model-based ARRs

> : XOR & : AND | : OR

**NNITI** 



				0-	25052	0	0	0	0	0	0	0	25
Algorithm	Scoring Time (s) 34276 samples	Accuracy (%)	f1-score (%)	1-	0	1301	500	0	0	0	0	0	- 20
DT4X $*$	2.48	84.37	82.09	2 -	0	1164	623	0	0	0	0	0	
sklDT	0.00	85.07	84.48	<u> </u>	0	0	0	0	0	0	0	996	- 15
m sklRF	0.17	84.95	83.72	lab	Ŭ	Ŭ	Ŭ	Ŭ	Ŭ	Ŭ	Ŭ	550	
sklLR	0.00	73.09	61.73	enul 4 -	0	0	0	0	0	102	0	1232	- 10
sklNB	0.02	16.80	20.56	5 -	0	0	0	0	0	530	0	0	
sklSVM	28.57	84.95	83.72	6-	0	0	0	0	0	537	0	506	- 50
m sklKNN	4.34	85.18	84.30							557		500	
				7 -	0	0	0	0	0	498	0	1235	
					1	1	1	1	1	1			- 0

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2

3

Predicted label

4

\* DT4X also finds the analytical expression of model based RRAs





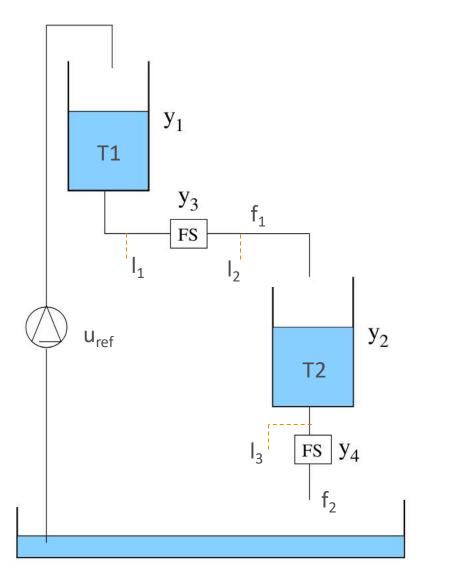
6 7

5

Static: 
$$x = (x_1, x_2, ..., x_n) \in \mathbb{R}^n$$
  
Dynamics:  $x = (x_1, ..., x_n, \dot{x}_1, ..., \dot{x}_n, ..., x_1^{(d)}, ..., x_n^{(d)}) \in \mathbb{R}^{(d+1)n}$ 



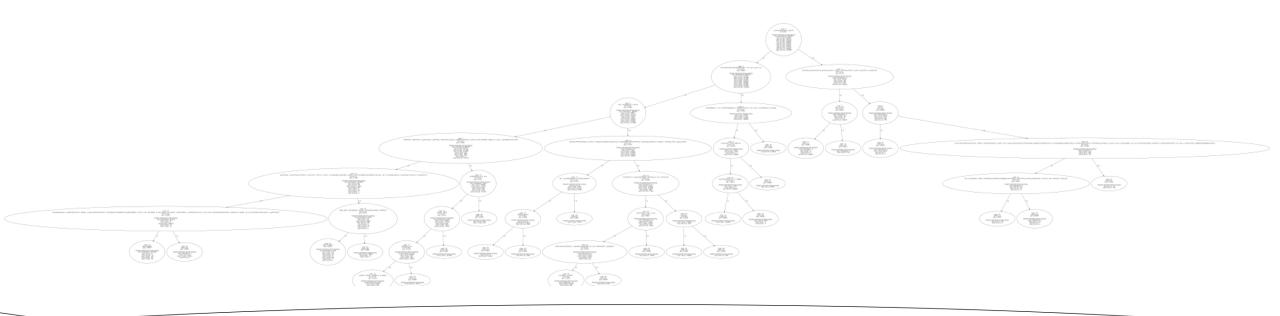




Type: dynamic Known RRAs: yes Data: simulated







node : 48 (d3/d\_y1 - d\_y2/d\_y3)\*(d5\*d6\*d\_y3\*\*2 + d\_y4\*\*0.5/d6)\*\*0.5\*(-(d\_y1 - d\_y4\*\*2 - d\_y4)\*\*2 + d\_y4\*\*0.25\*y4\*(-d6 + d\_y3)\*(d\_y3\*d\_y4 + d6/d2)\*\*1.0\*(d1\*d5\*d\_y3\*y4\*(d\_y1 - d\_y4\*\*0.5 - d\_y4)\*\*2)\*(d1\*d6\*d\_y3\*y4\*(d\_y1 + (d3\*\*2 + d\_y4\*\*0.5)\*\*0.5) - (d3\*\*2 + d\_y4\*\*0.5 - d\_y4\*\*2)\*\*2)\*\*2/d1) 8 vs 5 gini : 0.4978

 $\begin{array}{c} \text{Sample repartition during training:} \\ \text{class 0 (Faultless) : 10} \\ \text{class 5 (f-f2) : 316} \\ \text{class 8 (f-13) : 560} \\ \text{class 10 (f_-c2) : 22} \end{array}$ 



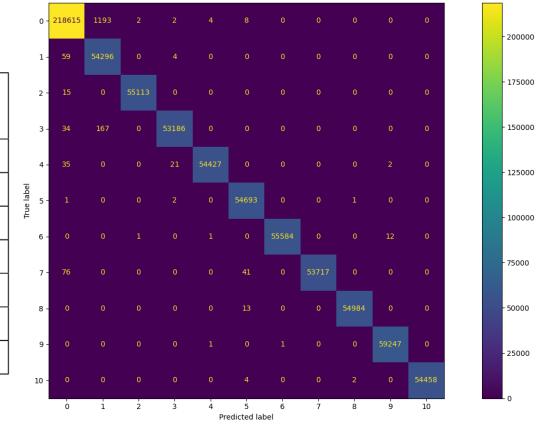


Algorithm	Scoring Time (s) 770022 samples	Accuracy (%)	f1-score $(\%)$
DT4X	373.74	99.78	99.78
sklDT	0.10	99.996	99.996
sklRF	7.30	99.997	99.997
sklLR	0.14	77.52	75.75
sklNB	0.49	77.51	78.37
sklSVM	-	-	-
sklKNN	15.10	99.998	99.998

Please note:

- DT4X does not find model-based ARR (syntactically)
- Isolation of classes structurally non-isolable

- Significantly slower predictions
- Performance close to the best algorithms



- **DT4X** is not only a data-based diagnosis system but it discovers the expressions of diagnosis indicators that provide explainability
- **Message**: do not forget the interpretable concepts and results of model-based methods !

#### **Ideas for future work**

- Boost symbolic classification with information extracted from the physical model of the system to reduce computation time and improve reliability (mostly for dynamic systems)
- Compare the genetic algorithm with **reinforcement learning**
- Analyse fault non-diagnosability in relation to **symbolic classification convergence** (not good f)



L. Goupil, L. Travé-Massuyès, E. Chanthery, T. Kohler, and S. Delautier. Tree based diagnosis enhanced with meta knowledge applied to dynamic systems. *Best Theory Paul M. Frank paper Award*, IFAC Safeprocess , June 2024, Ferrara, Italy. *IFAC-PapersOnLine*, 58(4):1–6, 2024.

#### Tree base diagnosis enhanced with meta knowledge

Diagnosis indicator discovery as a by product

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