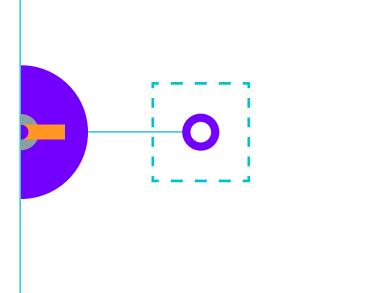


Agenda:

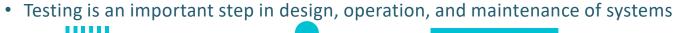


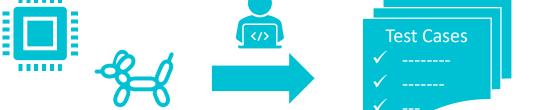
1. Motivation

- 2. Model-Based Testing
- 3. Testing with Decision Trees
- 4. Coverage Metric
- 5. Automatic Test Generation
- 6. Example
- 7. Conclusion

24.03.2023

Motivation



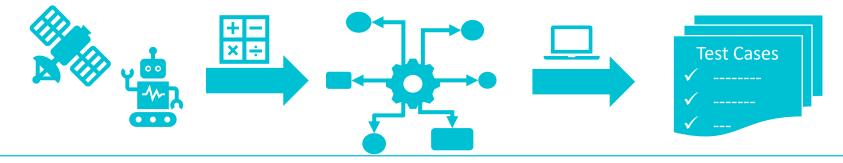


• For complex, black-box systems deriving test cases is particularly difficult

- → We propose a new MBT approach using decision tree models
- → Decision trees allow to learn from bounded history

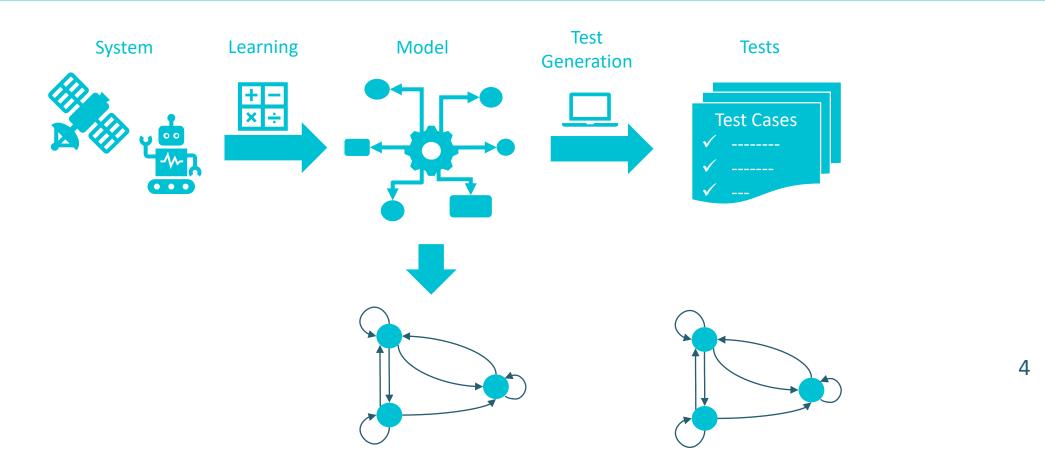


• A solution is model-based testing (MBT) with learned models



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Model-Based Testing



• MBT consists of a model learning and a test generation step. Often, finite automaton models are considered and state, transition, or other coverage criteria are used for test generation [1,2,3]

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Testing with Decision Trees

• Decision Tree Models represent the sequential behaviour of a system

 O_k

 l_k

D

• Feature vectors show N previous time steps and predict the output of the next time step

 l_{k+1}

i_t: *input symbol o_t*: *output symbol*

• In the following, we assume systems with a Mealy machine representation

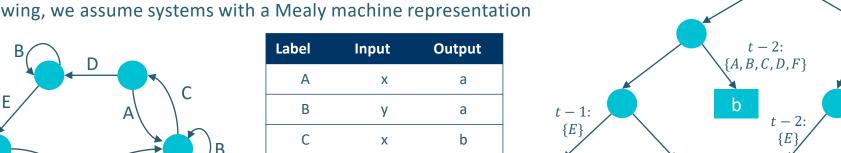
S. Plambeck, L. Schammer, and G. Fey, "On the viability of decision trees for learning models of systems," in Asia and South Pacific Design Automation Conference (ASP-DAC), 2022, pp. 696–701.

D

Е

F

Feature Vector



У

Х

y

b

С

С

 l_{k+2} O_{k+2}

 l_{k+N}

 O_{k+N}

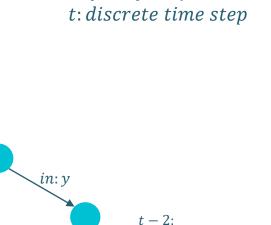
Class Label

а

in: x

t - 1:

 $\{C, F\}$



 $\{C, F\}$

n

а

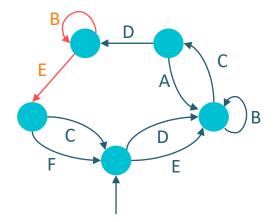
С

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Testing with Decision Trees

- The decision tree is learned from observations of bounded history
 - Enables model learning without knowledge of an initial state or possibility to return to the initial state
 - \rightarrow testing without reset to an initial state
- We call this Ad-hoc Testing



- Knowing the current history, we want to find future inputs to cover a maximum amount of system behaviour
- \rightarrow How to define coverage on a decision tree model?

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TUHH **Coverage Metric** • Most relevant system behaviour is encoded in the paths from root to leaf nodes • A discrete time step corresponds to an update of the current history... in: x in: y *t* – 2: $i_{k+2} o_{k+2}$ t - 2: i_k $i_{k+1} \ o_{k+1}$ $\{C,F\}$ *0*_{*k*} $\cdots i_{k+N}$ O_{k+N} $\{A, B, C, D, F\}$ b b *t* – 1: t - 2: $\{E\}$ $\{E\}$ а

[*E*, *C*, *x*] c

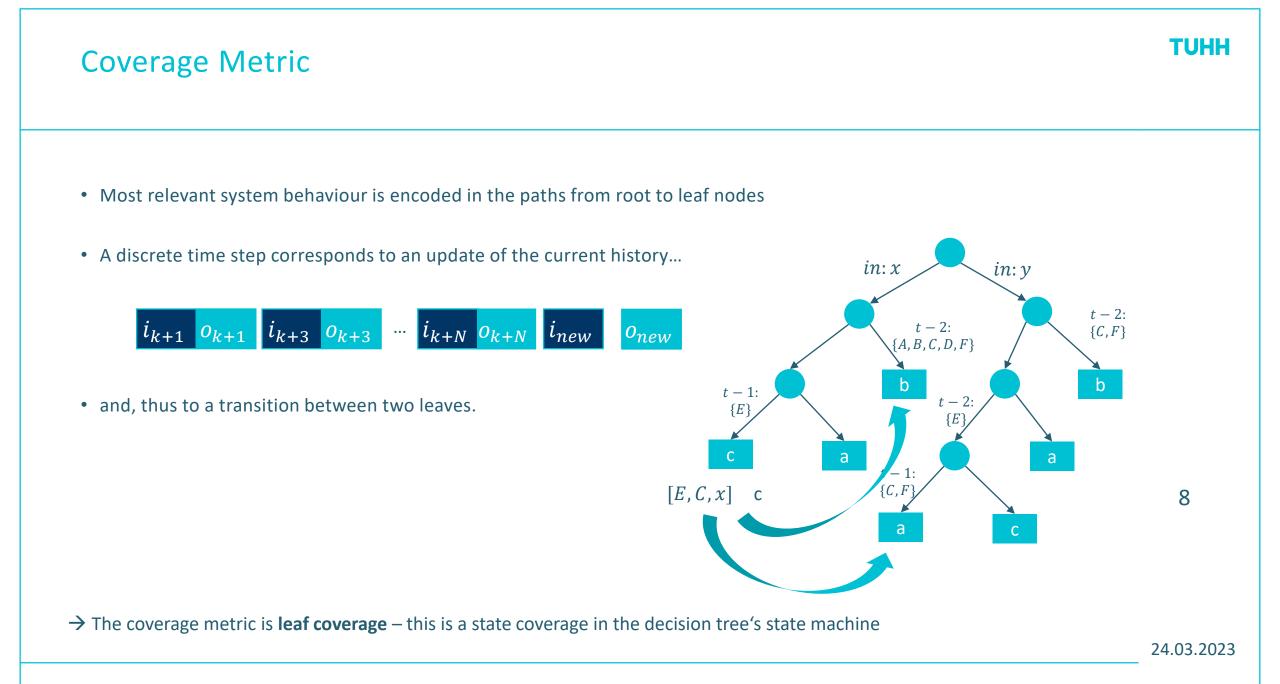
t - 1: {*C*, *F*}

а

C

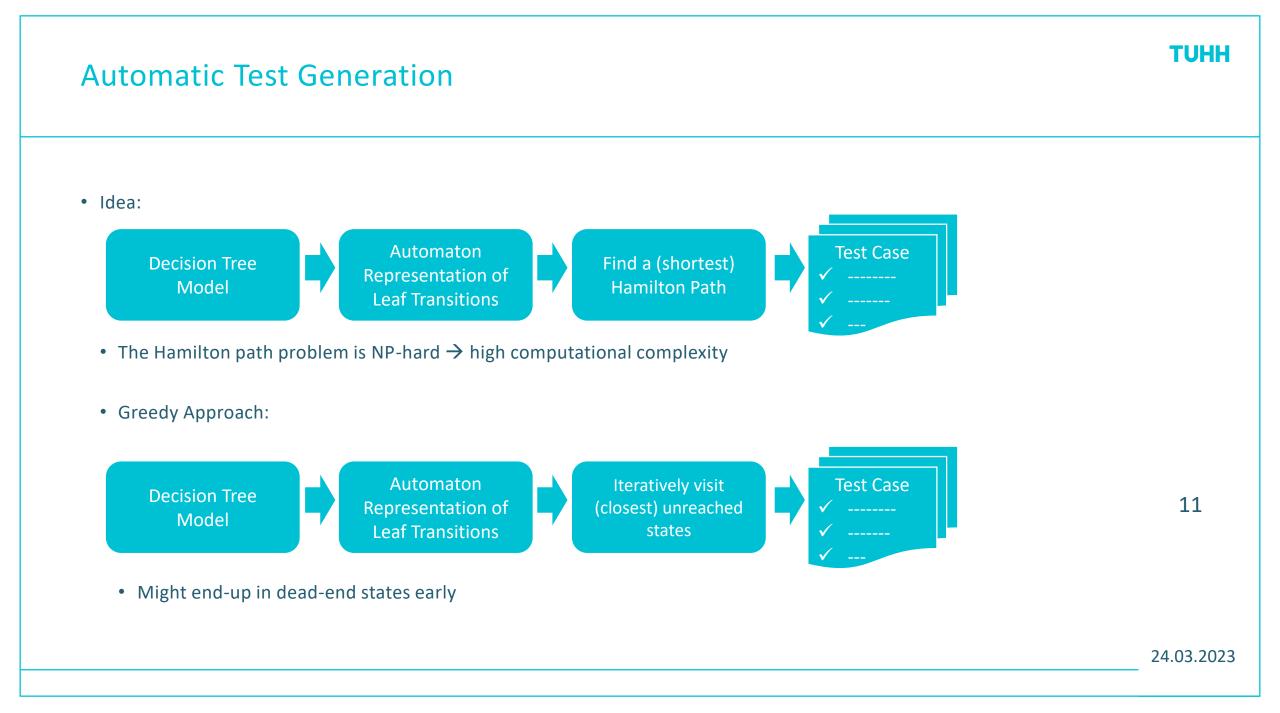
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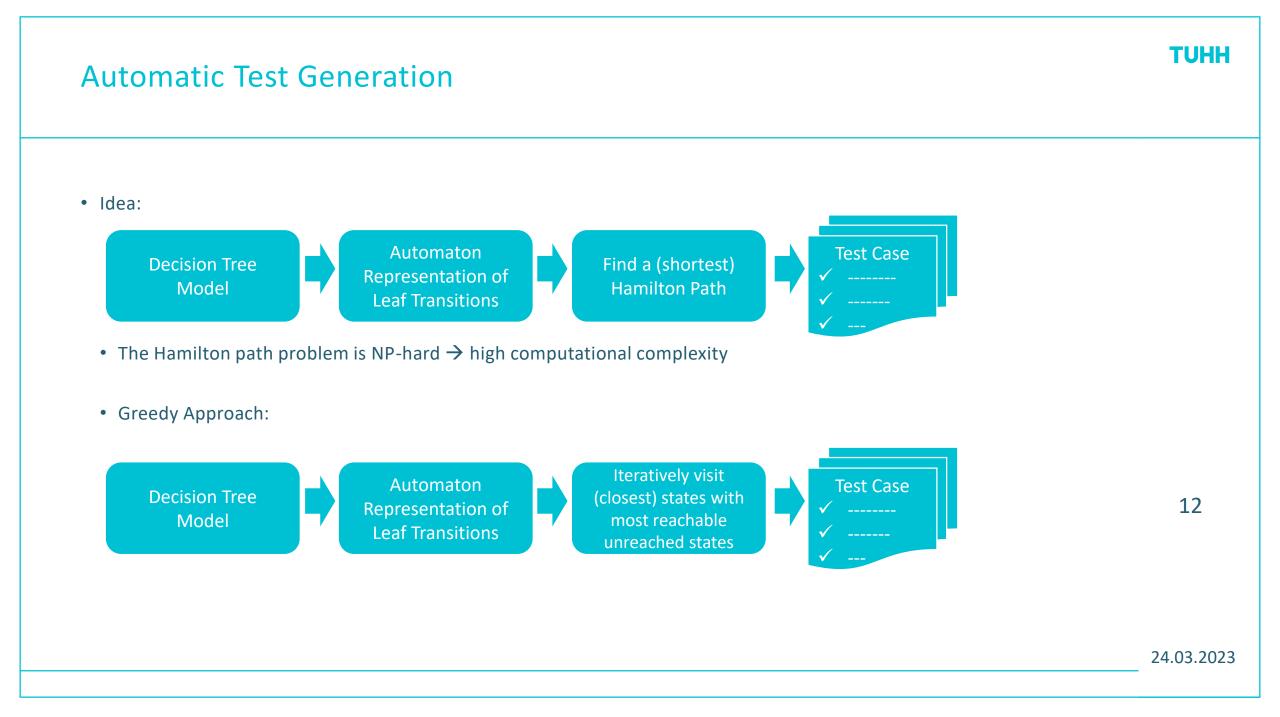
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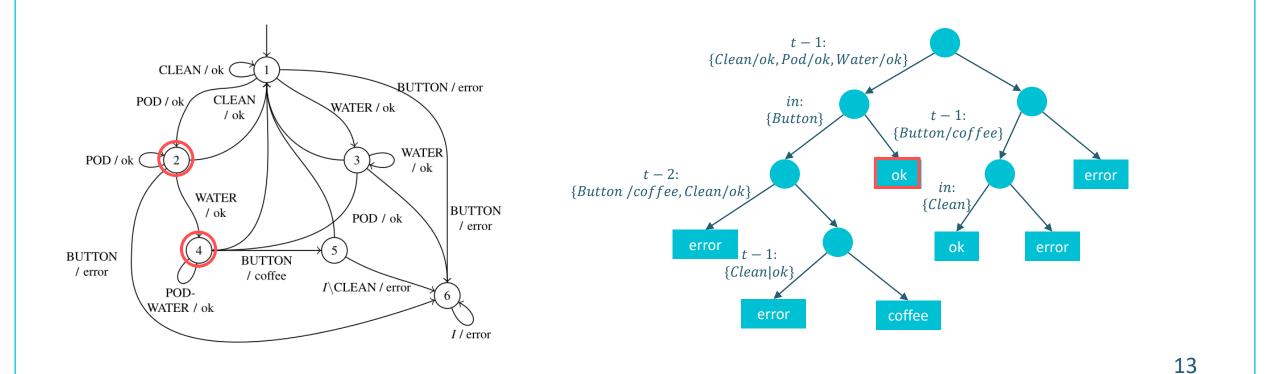


TUHH **Automatic Test Generation** • Idea: Automaton Test Case Find a path that Decision Tree Representation of visits all states Model Leaf Transitions 9 24.03.2023

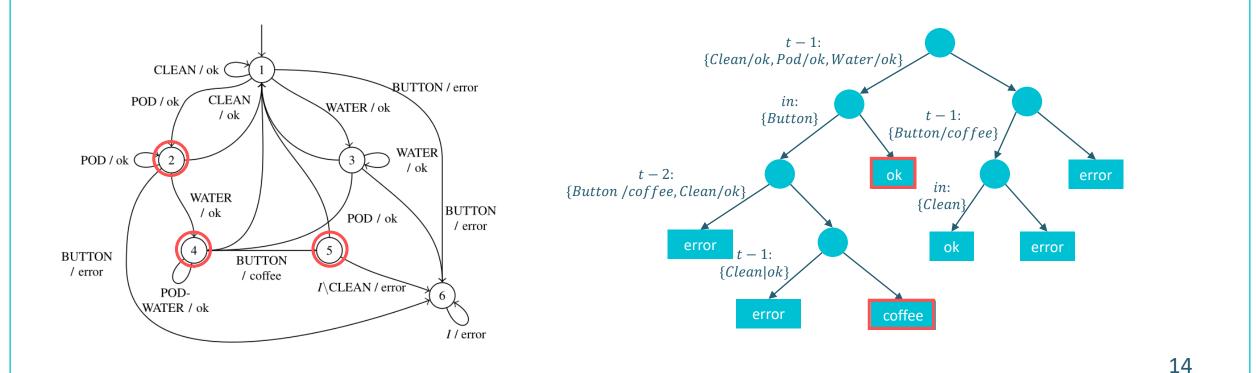
TUHH **Automatic Test Generation** • Idea: Automaton Test Case Find a (shortest) Decision Tree Representation of Hamilton Path Model Leaf Transitions • The Hamilton path problem is NP-hard \rightarrow high computational complexity 10 24.03.2023





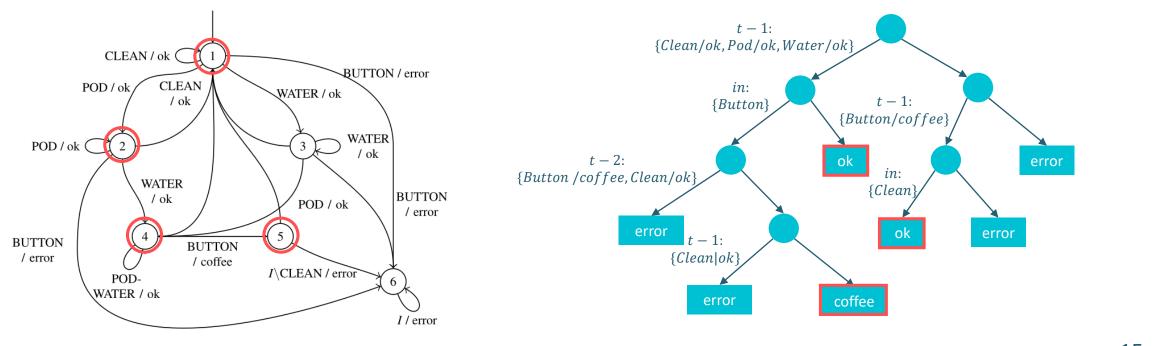


Assumption: current history is [*Clean/ok*, *Pod/ok*, *Water*] \rightarrow we are in state 2 and go to state 4 of the automaton



Assumption: current history is [Clean/ok, Pod/ok, Water]
→ we are in state 2 and go to state 4 of the automaton
→ The next output is ok

We choose a next input *Button* → The next history is [*Pod/ok*, *Water/ok*, *Button*]

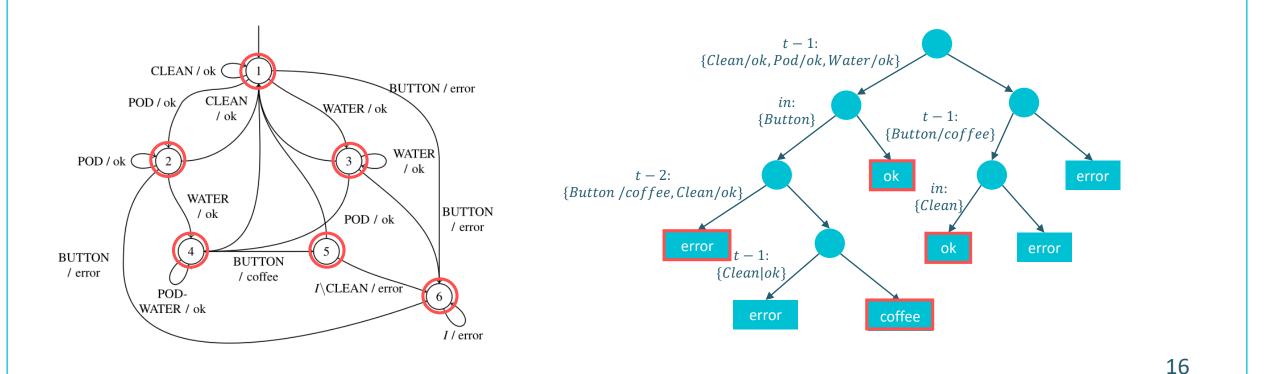


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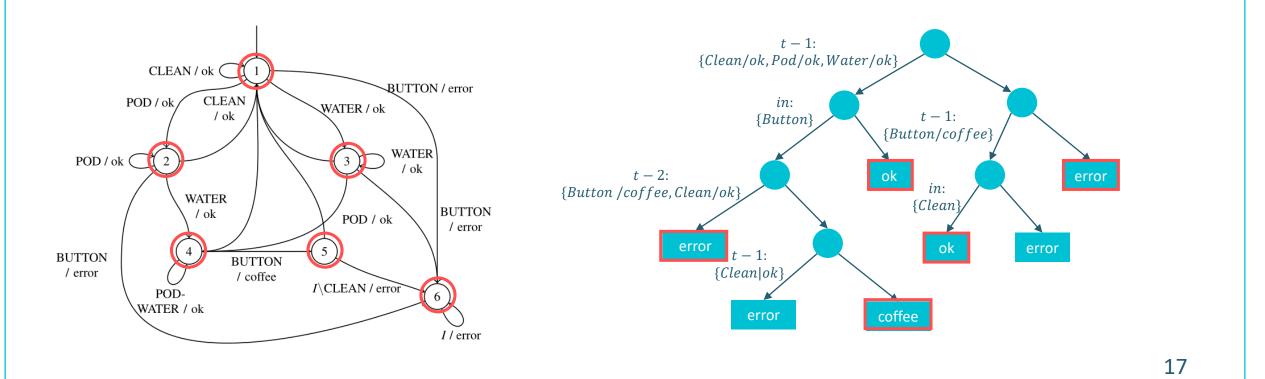
Assumption: current history is [Pod/ok, Water/ok, Button]
→ we are in state 4 and go to state 5 of the automaton
→ The next output is coffee

We choose a next input *Clean* → The next history is [*Water/ok*, *Button/coffee*, *Clean*]



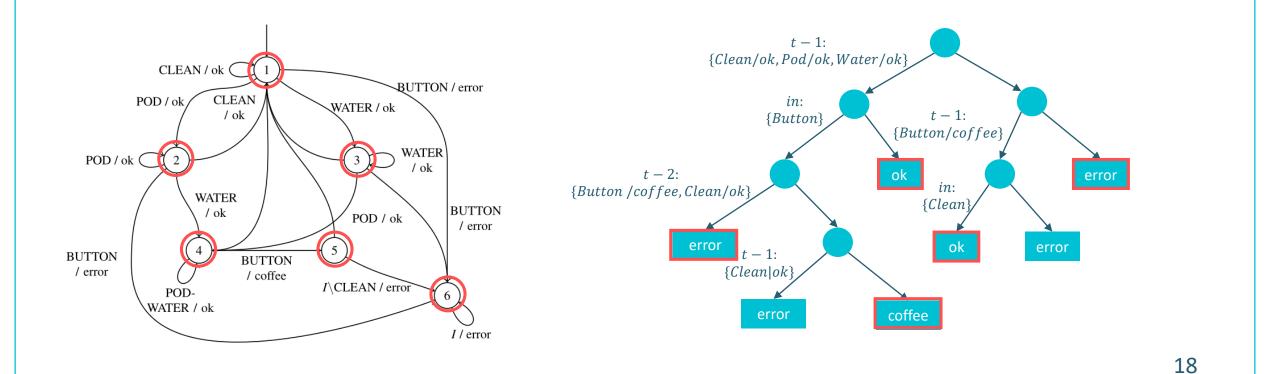
Assumption: current history is [*Water/ok*, *Button/coffee*, *Clean*] → we are in state 5 and go to state 1 of the automaton → The next output is *ok*

We choose a next input *Water*, *Button* → The next history is [*Clean/ok*, *Water/ok*, *Button*]



Assumption: current history is [Clean/ok, Water/ok, Button]
→ we are in state 3 and go to state 6 of the automaton
→ The next output is error

We choose a next input *Pod* → The next history is [*Water/ok*, *Button/error*, *Pod*]



 \rightarrow A leaf coverage of $\frac{5}{7}$ is reached while a full state coverage on the original automaton representation is achieved

Coffee Machine Example from: B. Steffen, F. Howar, and M. Merten, "Introduction to active automata learning from a practical perspective," in Formal Methods for Eternal Networked Software Systems, 2011, pp. 256–296.

Conclusion

- We introduced a new strategy for Model-Based Testing (MBT) using decision tree models
- The advantage is the learnability and, thus testability from bounded history (ad-hoc testing)
- We proposed multiple strategies to apply automatic test generation

- Future work considers
 - Comparison to existing MBT approaches
 - Evaluation of complexity and scalability

References

- [1] K. I. Eder, W. ling Huang, and J. Peleska, "Complete agent-driven model-based system testing for autonomous systems," in Workshop on Formal Methods for Autonomous Systems (FMAS), 2021.
- K. A. El-Fakih, R. Dorofeeva, N. V. Yevtushenko, and G. V. Bochmann, "Fsm-based testing from user defined faults adapted to incremental and mutation testing," Programming and Computer Software, vol. 38, no. 4, 2012.
- J. Peleska, E. Vorobev, and F. Lapschies, "Automated test case generation with smt-solving and abstract interpretation," in NASA Formal Methods. Berlin, Heidelberg: Springer Berlin Heidelberg, 2011, pp. 298–312.

Footnotes:

[4] S. Plambeck, L. Schammer, and G. Fey, "On the viability of decision trees for learning models of systems," in Asia and South Pacific Design Automation Conference (ASP-DAC), 2022, pp. 696–701.
 [5] B. Steffen, F. Howar, and M. Merten, "Introduction to active automata learning from a practical perspective,"

in Formal Methods for Eternal Networked Software Systems, 2011, pp. 256–296.

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Thank You

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