# Supporting Multi-View Models of Software Systems: Synthesis Techniques

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   State-based abstractions
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- Deductive LTS synthesis from guarded hMSC
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## Why Modeling Software Systems?

- Elaborating requirements and exploring system design
  - Reasoning about, verifying and documenting systems

## Why Multi-View Models?

- Different models => different but complementary focusses
- Example based or rule based?
- Agent interactions or agent internals?
  - Declarative or operational?

# How Modeling with Multi-View Models?

- Multi-view framework => inter-model consistency rules
  - Opportunity for synthesis-driven system modeling

## Multi-View Models: Golden Triangle

### Scenarios

Typical examples or counterexamples of system behavior through sequences of interactions among agents

Example-based, interactions, operational

#### Goals

Prescriptive statements of intent whose satisfaction requires cooperation among the agents forming the system

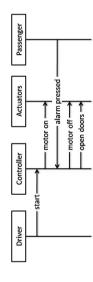
• Rule-based, interactions as well as agent internals, declarative

## State machines

Classes of required agent behaviors in terms of states and events firing transitions

Rule-based, agent internals, operational

# Example of synthesis-driven modeling



# Synthesize Controller's state machine

- Accepting at least the sequence of events shown in the example
   Under the control of descriptive properties: train doors are either opened or closed but not both in the same state
- Under the control of prescriptive properties: train doors must stay closed when the train moves

Background: Multi-View Formal Framework

- Event-based Behavior Models
  Scenarios as Message Sequence Charts (MSC)
  State machines as Labeled Transition Systems (LTS)

## State-based abstractions

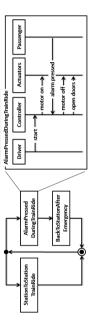
- System state through Fluents
   Guards in behavior models (g-LTS and g-hMSC)
  - Decorations on behavior models

- Goals as intentional models

  Goals and Fluent Linear Temporal Logic (FLTL)

  Linking FLTL and LTS: property and tester automata

# Message Sequence Charts: agent interaction examples



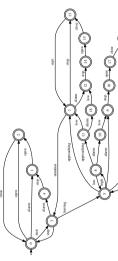
- MSC (right) and high-level MSC (left)

  Syntax of MSC and hMSC is described in [ITU96]

  Semantics of MSC and hMSC is defined in terms of Labeled Transition Systems, following [UKM03]

  We also allow a hMSC node to be refined as a finer-grained hMSC

Labeled Transition Systems (LTS) for agent behaviors



## **Labeled Transition Systems**

- Syntax and Semantics defined in [MK99]
   Each agent behavior is defined by a LTS. The system behavior is defined by LTS composition [MK99]
  - MSCs are admissible paths in the system LTS [UKM03]

# Capturing state information with Fluents

# Fluents capture the system state through the occurrence of

fluent  $Fl=<init_{Fl}$ ,  $term_{Fl}>initially\ Init_{Fl}$  where  $init_{Fl}$  and  $term_{Fl}$  are disjoint set of events rendering the fluent true and false, respectively

#### Example

fluent moving =< start,  $\{$  stop, emergency stop $\}>$  initially false fluent doors\_closed =< close,  $\{$  open, emergency open $\}>$  initially true

### A fluent is

- ... controlled by an agent if the agent controls (aka emits) all initiating and terminating events of the fluent
   ... monitored by an agent if the agent controls or monitors (aka receives) all initiating and terminating events of the fluent

#### Summary

**Guarded Behavior Models** 

Guards can be formally used in hMSC and LTS, leading to guarded hMSC (g-hMSC) and guarded LTS (g-LTS)

- A guard is a boolean expression on fluents
- Structured forms for hMSC and LTS, avoiding state/trace explosion
- Relax the assumption of fluent initial values being known for all instances

# Open question, i.e. not discussed in the paper

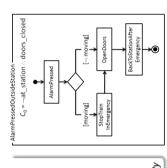
Architectural semantics of guards, i.e. what about guards and agents, guards and LTS composition, guard monitorability and controllability?

## Related publications

Damas C., Lambeau B., Roucoux F. and van Lamsweerde A., *Analyzing Critical Process Models through Behavior Model Synthesis*, in Proc. IGSE'2009: 31th International Conference on Software Engineering, Vancouver, Canada, May 16-24, 2009.

### Summary

- Decision nodes: outgoing transitions are labeled by boolean expressions on fluents
- Initial condition C<sub>0</sub> stating an invariant on the initial state
   Trace semantics through guarded LTS and LTS
   Automated checking of guards: non overlapping, completeness and reachability



### Summary

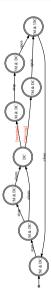
Guards in LTS, i.e. g-LTS

- A g-LTS transition is labeled by an event or a guard
   Initial condition C<sub>0</sub> stating an invariant on the initial state
   A trace is accepted by a g-LTS if three conditions hold: trace inclusion, admissible start and guard satisfaction

#### Example

#### Summary

#### Example



where M stands for moving and DC stands for doors\_closed

Goals expressed in FLTL

#### Goals

- Descriptive or prescriptive properties about the system [vL09]
  - Can be structured in AND/OR goal graphs

- Fluent Linear Temporal Logic (FLTL)

   Linear Temporal Logic (LTL, [MP92]) where propositions are Fluents
- FLTL is used by [GM03] to model check LTS against (state-based) temporal properties

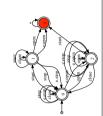
#### Example

- Maintain[DoorsClosedWhileMoving]: Train doors must always remain closed when the train is moving
   FormalDef: □(moving ⇒ doors\_closed)

- A Tester and Property LTS
  A Tester LTS can be synthesized from a FLTL safety property, as explained in [GM03]. The error state captures all event traces violating the property
  The Property LTS obtained by removing the error state captures all event traces not violating the property

#### Example

 $\bullet \ \Box(\textit{moving} \implies \textit{doors\_closed})$ 

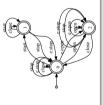


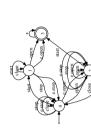
Aside: a regular language point of view on Tester and Property LTS

# Error state vs. accepting and non accepting states...

- Property (left) and Tester (right) automata are language complement of each other...
- The tester automaton (resp property) accepts all traces violating (resp. not violating) the safety property

#### Example





- Ouestion addressed

  What is the set of event traces accepted by a guarded hMSC?

  How to model-check a guarded model?

### Main results

- Adaptation of [UKM03] to synthesize a guarded LTS from a guarded hMSC
- LTS composition algorithm for synthesizing a pure LTS from a guarded LTS
  - Adaptation of [GM03] to model-check FLTL properties on guarded-LTS

Deductive LTS synthesis: what's done, what remains to be done?

## Work in progress

- Full state tracability through the g-hMSC, g-LTS, LTS synthesis (work in progress)
   Model-checker feedback in terms of the g-LTS and g-hMSC instead of the pure LTS

## Open questions

Only applied on the whole system so far, not with different agents in mind (lack of g-LTS composition/decomposition semantics)

## Related publications

Damas C., Lambeau B., Roucoux F. and van Lamsweerde A, Analyzing Critical Process Models through Behavior Model Synthesis, in Proc. ICSE'2009: 31th International Conference on Software Engineering, Vancouver, Canada, May 16:24, 2009.

# Inductive LTS synthesis from MSC and hMSC

## Limitation of deductive approaches

- Scenarios are known to be inherently partial, they only provide typical examples of system the usage
- Therefore, deductive techniques (e.g. [UKM03]) can only result in partial LTS
- Generalizing observed behaviors looks interesting in practice

## About our inductive approach

- Relies on Grammar Induction (GI), which provides a sound mathematical background
  - But how to
- Avoid poor behavior generalizations?
   Ensure consistency with other models (state variables, goals, ...)?
   Prune the induction process?

Short background on Grammar Induction

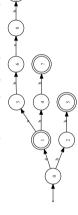
## In a few words

- Grammar induction aims at learning a regular language L from a set of positive and negative strings, i.e. respectively belonging and not belonging to the language
  - Also known as Automaton Induction when L is represented by a (Deterministic) Finite Automaton A(L)

## Known results ... among others

- RPNI algorithm (Regular Positive and Negative Inference)
- Necessary condition for convergence: structural completeness of the sample (each transition and accepting state of the automaton is used at least once in the input sample)
   Sufficient condition for convergence: the input sample contains a characteristic sample for A(L)

From a Prefix Tree Acceptor, accepting the positive sample only, ...



... induce a DFA by successively merging well-chosen state pairs (BlueFringe heuristics), under the control of the negative sample



# Grammar Induction for LTS Synthesis?

## Applicability

- A LTS is a DFA with all accepting states
   LTS define a subclass of regular languages (prefix-closed languages)
- A positive MSC is an accepting trace in the system LTS (obtained by composition of all agent LTS)
   A positive (resp. negative) MSC can be seen as a positive (resp. negative) string of the language accepted by the system LTS

#### The idea

- Learn the system LTS from positive and negative scenarios using RPNI
  - Synthesize agent LTS by projecting the system LTS on their monitored events (standard automaton algorithms)

# Grammar Induction for LTS Synthesis: a short summary

## **Question addressed**

- Importance of negative strings; are they initially available from end-users?
- How to ensure consistency with other models?
- Assumption that all MSC start in the same system state, what about loops and reuse?

### Main results

- Our interactive variant of RPNI, namely QSM, when few negative scenarios are initially provided
- Injecting fluent definitions, goals and legacy components ensures inter-model consistency and prunes the process
  - Other contributed variants, namely ASM and ASM\*, support an hMSC as input, relaxing the assumption mentionned

Interactive LTS Synthesis from MSC: the QSM algorithm Interactive LTS Synthesis from MSC

#### Summary

- RPNI with BlueFringe heuristic for selecting state pairs to merge
- When two states are merged, scenario queries are submitted to an end-user for classification as positive or negative behaviors
- The end-user, aka Oracle, guides the induction process, avoiding poor generalizations
  - Query generation relies on the definition of a characteristic sample, which provides a convergence criteria

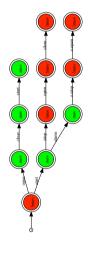
## Related publications

- Damas C., Lambeau B., and van Lamsweerde A, Generating Annotated Behavior Models
  From End-User Scenarios, IEEE Transactions on Software Engineering, Special Issue on
  Infraraction and State-based Modeling, Vol. 31, No. 12, pp. 1056-1073, 2005.
   P. Dupont, B. Lambeau, C. Damas, and A. van Lamsweerde, The QSM Algorithm and its
  Application to Software Behavior Model Induction, Applied Artificial Intelligence, Vol. 22,
  2008, 77-115.

### Summary

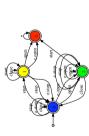
- PTA states can be decorated with fluent values, using the decoration algorithm
   The induction process can be constrained to avoid merging non equivalent states

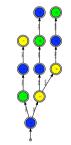
 $\textit{fluent mov(ing)} = < \textit{start}, \{\textit{stop}, \textit{emergency stop}\} > \textit{initially false}$ 



# Pruning the induction with goals

Summary
Color PTA states with corresponding states in the tester. Avoid merging two PTA states not sharing the color





Open questions

Sound but too strong! This is related to another open issue on ASM\* about the negative language L<sup>-</sup> (see later)

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# Pruning the induction though equivalence classes

## State coloring as a generalization

- Equivalence relations can be defined on PTA states and induction process constrained to avoid merging non equivalent states.
   Legacy components can be used in a similar way, for example

## Open questions

- Defining equivalence classes could be extended in the light of lattice-based decorations
- We are convinced that additional architectural constraints could be used (to avoid introducing implied scenarios, for example)

 C. Damas, B. Lambeau and A. van Lamsweerde, Scenarios, Goals, and State Machines: a Win-Win Partnership for Model Synthesis, Proc. FSE'06: Intl. ACM Symposium on the Foundations of Software Engineering, Portland (OR), November 2006. Related publications

Pruning the induction with control information

## Questions addressed

- Assumption that all MSCs start in the same initial state
  - End-users would like to capture loops when known
- Would it be possible to take a hMSC as input instead of a collection of MSCs?

### Main results

- Introduction of mandatory merge constraints on PTA states, which is the logical counterpart of equivalence classes
  - The ASM algorithm generalizes a positive language L<sup>+</sup> under the control of a negative sample S<sup>-</sup>. ASM\* generalizes a positive language L<sup>+</sup> under the control of a negative language L<sup>-</sup>
     Therefore, behaviors described in a hMSC can be generalized under the control of negative MSCs

# Pruning the induction with control information

## Open questions

- $\bullet$  Three concurrent techniques about pruning with goals: equivalent classes (too strong),  $ASM^*$  (looks not applicable directly) and Coste's work in [CFKdIH04]
  - ASM and ASM\* do not support the BlueFringe heuristics nor the interactive feature
     Theoretical GI questions because ASM and ASM\* do not perfectly fit the classical regular language learning framework

## Related publications

- B. Lambeau, C. Damas and P. Dupont, *State-merging DFA Induction Algorithms with Mandatory Merge Constraints*, Lecture Notes in Artificial Intelligence No. 5278, Springer, pp. 139-153, 2008, 9th International Colloquium on Grammatical Inference, St Malo, France, September 22-24.

   No feedback in the RE community so far

### Evaluation

## Deductive synthesis evaluation

- Usage of the g-hMSC model-checker is illustrated on a case study in [DLRvL09]
  - Additional case-studies: work in progress

## Inductive synthesis evaluation

- Number of generated queries and convergence of QSM have been evaluated on RE case studies and synthetic data in [DLDvL05] and [DLDvL08]
- Evaluation of the pruning techniques on RE case studies in [DLvL06]
   Evaluation protocol for ASM in [BLD08], applied on one RE case study and synthetic data

## **Tool Support**

- Induction Toolkit
  Recent refactoring of all the automaton tools (still in progress)
  Light release will be provided during the Stamina induction competition (2011)
  - Full release will be provided after the competition (few work remaining here)

# A FLTL model-checker for g-hMSC models

- Work in progress (architecture and packaging of the tool support)
   Future collaboration with Antoine Cailliau for additional (F)LTL tools (master thesis)

### References

- G. Damas B. Lambeau and P. Dupont
  State-merging de induction apportrins with nandatory merge constraints.
  In Grammatical Inserves, IGST08, number 5278 in Lacture Notes in Artificial
  - F Coste, D. Fredoulle, C. Kermorvant, and C. de la Higuera Introducing domain and typing bias in automata inference. In Greenmatical Interence. Altorithms and Applications, put
- C. Damas, B. Lambeau, P. Dupont, and A. van Lamsweerde.
  Generating annotated behavior models from end-user scenarios.
  IEEE Transactions on Software Engineering, 31(12):1056–1073,
- P. Dupont, B. Lambeau, C. Damas, and A. van Lamsweerde.
  The QSM algorithm and its application to software behavior model induction.
  Applied Artificial Intelligence, 22:77–115, 2008.
  - Dames, B. Lambeau, F. Roucoux, and A. van Lameweede.
    Analyzing critical process modes through behavior model synthering in ICSE DB. 31th International Conference on Software Engine

- C. Domms. B. Lamboau, and A. van Lamoweerdo.
  Senarios, goals, and state machines: a win-win partnership for model synthesis.
  In international ACM Symposum on the Foundations of Schewer Engineering, pa

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