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Towards Engineering Smart Cyber-Physical Systems with Graph Transformation

Systems

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Outline



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- 1. Motivation
- 2. Being Smart
- 3. Smart & Learning
- 4. Conclusions & Outlook

Outline



1. Motivation

- 2. Being Smart
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1. Motivation The Future: You name it ...

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A Selection of Critical Future Challenges (1/5)



Operational and managerial independence

- operated independent from each other without global coordination
- no centralized management decisions (possibly confliction decisions)
- decoupled evolution and co-existence of different versions
- Dynamic architecture and openness
 - must be able to dynamically adapt/absorb structural deviations
 - subsystems may join or leave over time in a not pre-planned manner

Challenge: Service-Oriented Architecture





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Service-Oriented Architecture (Service oriented architecture Modeling Language (SoaML)):



- Dedicated services are offered by systems via defined service contracts can be offered, looked up, and bound at run-time
- Interoperability is provided by a service bus
- a UML profile for modeling that supports collaborations as first class elements (service contracts) and links collaborations with component-based models

Observations:

- Service contracts permit to realize operational and managerial independence
- Offering, look up, and bind service and runtime supports dynamic architectures and openness
- But: dynamics not covered and
 no support for real-time or physics

A Selection of Critical Future Challenges (2/5)



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Advanced adaptation

Challenge: Advanced Adaptation



"Adaptation is needed to compensate for changes in the mission requirements [...] and operating environments [...]" [Northrop+2006]

"The vision of Cyber-Physical System (CPS) is that of open, ubiquitous systems of coordinated computing and physical elements which interactively **adapt to their context, are capable of learning, dynamically and automatically reconfigure themselves ...**" [Broy+2012]

"Incrementality comes into play because often changes are local to restricted parts. [...] **Incrementality** becomes even more necessary when changes occur at runtime and the software itself is responsible for reacting in a **selfmanaged manner**. In this setting, the processing that needs to be performed after each change is subject to severe time constraints."

[Ghezzi2012]

Required kind of self-adaptation:

System level adaptation

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- System-of-systems level self-adaptation
- Incremental self-adaptation

A Selection of Critical Future Challenges (3/5)



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 - subsystems may join or leave over time in a not pre-planned manner
- Advanced adaptation
 - Resilience

Challenge: Resilience

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"The vision of Cyber-Physical System (CPS) is that of open, ubiquitous systems [...] which [...] and **fulfill stringent safety, security and private data protection regulations**." [Broy+2012]

"Resilience[:] This area is the attribute of a system, in this case a SoS that makes it less likely to experience failure and more likely to recover from a major disruption." [Valerdi+2008]

"Resilience is the capability of a system with specific characteristics before, during and after a disruption to absorb the disruption, recover to an acceptable level of performance, and sustain that level for an acceptable period of time." Resilient Systems Working Group, INCOSE

Required coverage of resilience:

- Physical and control elements (via layers of idealization)
- Software elements (via layers of abstraction)
- Horizontal and vertical composition of layers

A Selection of Critical Future Challenges (4/5)



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- Advanced adaptation
- Resilience
- Cross-Domain Integration

Challenge: Cross-Domain Integration



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Example: A convoy of fully autonomous cars abandons the premium track in order to give way to an ambulance (intersection of CPS specific for **traffic** and **health care**)

CPS of different domains have to be connected:



- According to social and spatial network topologies, CPS operate across different nested spheres of uncertainty
- CPS dedicated to different domains have to to interact and coordinate.

Integration has to cover multiple domains and their paradigms

A Selection of Critical Future Challenges (5/5)



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- Operational and managerial independence
 - operated independent from each other without global coordination
 - no centralized management decisions (possibly confliction decisions)
 - decoupled evolution and co-existence of different versions
- **Dynamic architecture** and **openness**
 - must be able to dynamically adapt/absorb structural deviations
 - subsystems may join or leave over time in a not pre-planned manner
- Advanced adaptation
- Resilience
- Cross-Domain Integration
- Integrate Models of Computation

Challenge: Integrate Models of Computation

- Problem to integrate models within one layer as different models of computation are employed
 - Leaky abstractions are caused by lack of composability across system layers. Consequences:
 - intractable interactions
 - unpredictable system level behavior
 - full-system verification does not scale



Integration has to cover multiple layers and their paradigms



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2. Being Smart: Adaptation Feedback Loop (MAPE-K)



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Awareness Enough?

Simple Self-Awareness : Let's have a look at Nature ...



Ant colonies operate as a superorganism that

combines information processing of many ants and their interaction with the environment at the physical level (using stigmergy as coordination mechanism).

Example:

Asymmetric binary bridge experiment

Observations:

- Initially both options will be taken with the same probability.
- The concentration of the pheromones will increase faster on the shorter path.
- The higher concentration of pheromones on the shorter path will make it more likely that an ant choses this shorter one.
- Positive feedback will amplify this effect and thus finally the longer path will only be used seldom.
- → Can our problems be solved by borrow ideas from nature?



Simple Self-Awareness : Let's have a look at Nature ...



Another Example:

"Ant Mill"

Observations:

Such a behavior would be not acceptable for an engineered system even for unexpected circumstances (rare events).



- If even "Nature" come up with designed solutions that fail (even evolution selected for ages), how could we envision to be more successful?
- But there is also a solution in nature:

reflect on itself (None Simple Self-Awareness)

Complex Aware-ness: Runtime Models



Simple Awareness "without" models:

 Still explicit or implicit design-time models are used guide adaptation processes

Complex Awareness with runtime models:

- Explicit runtime models are learned to guide adaptation processes and allow reasoning
- Model as reference can include goals
- Model of software + context capture changes
- Limitation: covers only changes captured by the runtime models (possibly multiple!); requires correct adjustment from observations
- Option: adaptation can also reason with these models (e.g., predict outcome of changes)



Self-Awareness & System of Systems





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Self-Adaptive Systems:

- Make systems self-aware, context-aware, and requirements-aware using some form of reflection
- Enable systems to adjust their structure/behavior accordingly

Self-Organization:

 The capability of a system of systems to organize their structure/behavior without a central control (emergent behavior)

Observations:

- a spectrum from centralized top-down self-adaptation to decentralized bottom-up self-organization with many intermediate forms exists
- existing (formal) models and analysis approaches for CPS are no longer applicable as they do not cover reflection/adaptation (design, verification, ...)

Some Related Observations ...





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- Service-Oriented Architecture can be described by a graph of links between the systems that evolve
- Self-Adaptive and Self-Organization can be described by a graph of links between the components resp. systems that evolve/reconfigure and in case of reflection most models can be described by such a graph as well
- Runtime Models can be described by a dynamic graph of models and links between them

Graph transformation systems encoding models and their linking would allow to combine Service-Oriented Architecture, Self-Adaptive / Self-Organization, and Runtime Models with evolving structures and **could be** the basis for a **solid foundation ...**

Complex Self-Awareness: A look at a complex example...



A system of **autonomous shuttles** that operate on demand and in a decentralized manner using a **wireless network**.



Related Observation Concerning the Example



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Modeling Problems:

Shuttles move on a topology of tracks

HPI

Hasso

Plattner Institut

non-functiona

Arbitrary large topologies

Solution:

- State = Graph
- Reconfiguration rules = graph transformation rules
- Safety properties = forbidden graphs
- Formal Verification possible

Very strong reduction: not all properties are represented

- Dynamic convoy structures and movement of the shuttles on the topology of tracks
- Real-Time movement of the shuttles on the topology of tracks
- Real-Time protocols for convoy coordination
- Continuous driving behavior
- Random communication errors

Graph Transformation Systems: Naïve Example



- Map the tracks
 - Map the shuttles



 Map the movement to rules (movement equals dynamic structural adaptation on the abstract level)



Graph Transformation Systems: Naïve Example





Runtime Models & Idealized Perception

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[Giese+2015]

SMARTSOS suggests:

 use a graph of links between the systems, components, and

internal represented data as well as

 use graph transformations to capture possible changes

to model

- Service-Oriented Architecture,
- Self-Adaptive and Self-Organization, and
- Runtime Models

Idealized Consistent Cyber & Physical World



[Giese+2015]



Runtime Models & HPI Hasso Exchanging Perceptions



[Giese+2015]







Related Observation Concerning the Example



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Coverage of the Challenges for Models

Needs:

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- Operational and managerial independence
- Dynamic architecture and openness
- Scale for local systems or networked resp. large-scale systems of systems
- Integration of the physical, cyber, (and social) dimension
- Adaptation at the system and system of system level
- Decoupled evolution and co-existence of different versions
- Resilience of the system of system



HPI

Hasso

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Coverage of the Challenges for Analysis



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Needs:

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- **Model Characteristics:**
- Compositionality
- Dynamic structures ²
- Abstraction
- Hybrid behavior
- Non-deterministic
- Reflection for models
- Incremental extensions

Probabilistic

Our Work:

SMARTSOS

 Checking Inductive Invariants for GTS ([Becker+2006]), Timed GTS ([Becker&Giese2008]), and Hybrid GTS ([Becker&Giese2012]) and Checking k-Inductive Invariants for GTS ([Dyck&Giese2017]) resp. Attributed GTS ([Schneider+2020])

- Model Checking Probabilistic
 GTS ([Krause&Giese2012])
- Probabilistic timed GTS: Model Checking ([Maximova+2018), Compositional Model Checking ([Maximova+2021])

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3. Smart & Learning: Learning vs. Training





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Learning allows to adjust the behavior of systems:

Trained systems:

- learning only offline
 - BUT: additional surveillance must be online
- Learning systems:
 - Often initially trained
 - Steady improvement
 by learning **online**
 - additional surveillance needed?

Simple Awareness & Learning



Train goals: adjust goals according to success w.r.t. higher level goals

Variable goals: react to changes of the goals

Static goals:

no variation



No learning: react to perceptions directly; avoid explicit model at run-time

Training/Learn perception: how to interpret observations

Complex Self-Awareness & Learning



Train/Learn goals:

Train/Learn adaptation: ...

adjust goals according to success w.r.t. higher level goals

Variable goals: react to changes of the goals

Static goals:

no variation



Learn also runtime model concepts (unknown unknowns); runtime models evolve according to the perception of useful differences

Learn runtime models (known unknowns); parameters, elements, and relations of runtime models are **learned** according to the perception

No learning: react to perceptions directly; avoid explicit model at run-time

Train/Learn perception: how to interpret observations

Complex Self-Awareness & Learn Context (1/3)





Server (Registry of the section control; not global!):

Offers track profile (distributed learning of a runtime model of the track)

Client (Monitor of the shuttle):

- Applies track profile (local learning of a runtime model of the shuttle and planning an adaptation in form of an optimal trajectory)
- Must handle cases where the service is available or not

Complex Self-Awareness & Learn Context (2/3)

Scheme:

server



Suspension/tilt module

- □ air springs (filter for higher frequencies)
- □ active suspension system (lower frequencies)

We consider three different control strategies:

- (1) robust controller: track as reference point; damping the relative movement ⇒ only achieves moderate damping.
- (2) absolute controller uses a virtual skyhook in order to ensure the absolute acceleration of the shuttle body is minimized
 ⇒ comfort usually maximized; problematic on inclines
- (3) reference controller: Instead of virtual skyhook, the real track is used as reference ⇒ highest comfort; requires data about the track

Client proxy:

network

Find local responsible registry

:client proxy

register at the local registry (requestInfo)

:mode mar

:control

version 2

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modes

(events; discrete)

control

(signals; continuous)

- Receive data from the registry (sendInfo)
- Manage cases where the data is available or not (outside the proxy)
- Send data to the registry (experience)
- PLUS: detect invalid runtime model!

[Burmester+2008]

Complex Self-Awareness & Learn Context (3/3)



Distributed learning runtime models

(known unknowns); parameters, elements, and relations of runtime models are learned according to the perception of other agents



Static goals:

Learning runtime models

(known unknowns); parameters, elements, and relations of runtime models are **learned** according to the perception

OBSERVATION: There is no guarantee that the runtime models are not invalid due to fact that they always rely on potentially erroneous or outdated measurements/ perceptions → detection + backup strategy always necessary

Complex Self-Awareness & Train Goals (1/4)





Required: Function computing the impact on the utility for each possible rule application **Open Question:** Can we learn these functions offline (training)?

Complex Self-Awareness & Train Goals (2/4)





Complex Self-Awareness & Train Goals (3/4)



[Ghahremani+2018]

YES

RQ: Does the performance approximate the analytic-defined optimum?



Normalized rewards across prediction models for the combined variant

Normalized Reward (mod)= $\frac{Reward (mod)-Reward(Baseline)}{Reward (Optimal)-Reward(Baseline)}$

Complex Self-Awareness & Train Goals (4/4)



Train goals: adjust goals according to success w.r.t. higher level goals **PROBLEM:** There is no guarantee that the trained goals are valid due to fact that they always rely on potentially erroneous or outdated measurements/perceptions
→ optimality is not guaranteed

Learn runtime models (known unknowns); parameters, elements, and relations of runtime models are learned according to the perception





Outline



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4. Conclusions & Outlook



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 Graph transformation systems encoding models and their linking allow to combine Service-Oriented Architecture, Self-Adaptive / Self-Organization, and Runtime Models with evolving structures and are a suitable basis for a solid foundation.
 - Runtime models and via collaborations shared runtime models enabled Self-Adaptation and Self-Organization of the systems and system of system
 - Limitations:
 - Model is a rather strong idealization
 - Analysis relies on the validity/trustworthiness of the models
 - Development-time models may become invalid over time
 - Runtime models for self-awareness may become invalid

Outlook (1/5) Modeling



Needs:

- Operational and managerial independence
- Dynamic architecture and openness
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Outlook (2/5) Analysis



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Needs:

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Model Characteristics:

SMARTSOS

- Compositionality
- Dynamic structures
- Abstraction
- Hybrid behavior
- Non-deterministic
- Reflection for models
- Incremental extensions

Probabilistic

BUT: We have to assure resilience for complex sequence properties (even ensemble properties) of **hybrid probabilistic infinite state systems**.

AND: cover the learning ...

State-of-the-Art & our Work:

 Checking Inductive Invariants for GTS ([Becker+2006]), Timed GTS ([Becker&Giese2008]), and Hybrid GTS ([Becker&Giese2012]) and Checking k-Inductive Invariants for GTS ([Dyck&Giese2017]) resp. Attributed GTS ([Schneider+2020])

Only state properties!

Model Checking Hybrid Systems Only sequence properties for finite

discrete state systems with rather bad scalability!

- Model Checking Probabilistic
 GTS ([Krause&Giese2012])
- Probabilistic timed GTS: Model Checking ([Maximova+2018), Compositional Model Checking ([Maximova+2021])

Only very restricted probabilistic sequence properties ...

Outlook (3/5)



Any approaches supporting Service-Oriented Architecture, Self-Adaptive / Self-Organization, and Runtime Models with evolving structures and Training/Learning will face tough challenges:

Challenges:

- Models have to capture many characteristics (e.g., dynamic structure, hybrid) usually tackled using different paradigms
- Even in case of strong idealization often relevant properties can not be analyzed
- Analysis relies on the validity/trustworthiness of the models
 - Development-time models may become invalid over time
 - Runtime models for self-awareness may become invalid

Outlook (4/5)

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Runtime models for self-awareness may become invalid



Learn also runtime model concepts (unknown unknowns); runtime models evolve according to the perception of useful differences

Learn runtime models (known unknowns); parameters, elements, and relations of runtime models are learned according to the perception

Train/Learn perception: how to interpret observations

Correctness of the training/learning, is crucial if we have no **robust backup** and guarantees are required

Outlook (5/5)



⁵⁰ Runtime models for self-awareness may become invalid **BUT:** Development-time models will become invalid over time

Static goals: no variation



No learning: react to perceptions directly; avoid explicit model at run-time

Avoiding training/learning is also not safe (in the long run) if we do not have a **robust solution**!

Some Literature



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Books:

- Software Engineering for Self-Adaptive Systems
- Software Engineering for Self-Adaptive Systems II
- Software Engineering for Self-Adaptive Systems III. Assurances
- Self-Aware Computing Systems

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