

Joint analysis of safety and security

Identifying interactions between safety and security, from model level
to language level

Lada Egorova
CTSIS team, LCIS

Supervised by: Oum-El-Kheir Aktouf
School Tutor: Om Essaad Slama
The chairman of the jury: Yann Kieffer

24 June 2025

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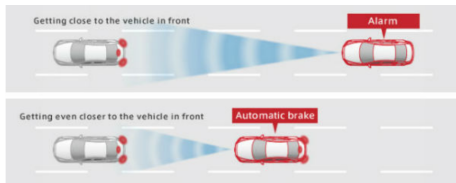


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Problem context

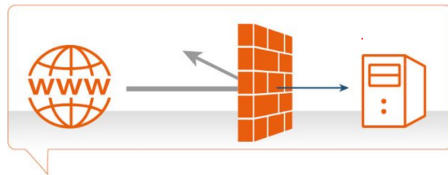
Auto-braking - Safety feature

In critical situation, system tries to send a message to stop the vehicle



Firewall - Security feature

Firewall blocks the unexpected messages, which can lead to a danger



Problem context

Finding interactions can be difficult:

- Requires manual analysis
- Developers of safety and security features usually don't interact with each other
- The interactions are hard to predict



Existing methods

Approach Type	Method / Tool	Application / Notes
Game-Theoretical	Game theory	Used for co-verification and cyberspace conflict modeling
Tree-Level Models	Fear Event Trees, Multi-Level Models, Fault Trees	Analyze origins of safety-security interactions
Knowledge Graphs	Dual-security analysis	Applied to ICS for enhanced threat modeling and traceability
Formal Methods	Event-B framework	Formal guarantees; limited scalability
Matrix-Based	Inspired by Analytical Network Process (ANP)	Combined risk assessment in cyber-physical systems
AI / ML Based	Large Language Models (LLMs)	Has potential for safety-security management
Bowtie Analysis	Bowtie diagrams	Visual risk assessments in Industrial Control Systems
Other Tools	Semi-Automatic Tools, Standards & Co-Engineering, Bowtie Analysis	Used for specific cases

LCIS lab was founded in October 1996 and now brings together:

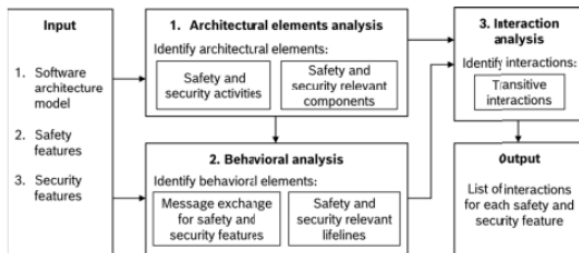
- 26 teacher-researchers
- 20 to 30 Ph.D. students
- postdoctoral researchers and research collaborators
- around 20 interns



Problem context: FISS

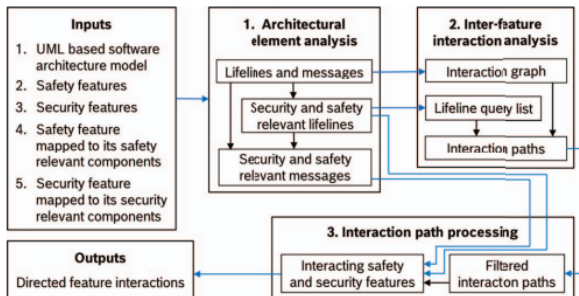
In LCIS lab, FISS (and its improved version, I-FASST), were developed:

- The tool takes a UML sequence diagram as an XML file as an input
- Returns a list of existing interactions between safety and security features
- Works iteratively for each feature



Problem context: I-FASST

- Has more inputs
- Filters paths with interaction chains
- Runs in one go



Problem context: possible contributions

Solution based on more general model (e.g., knowledge graphs)	Runtime verification research
<ul style="list-style-type: none">• XML files are not easy to edit in runtime• The general model solution does not depend on concrete realization• A general model may allow us to store different types of information	<ul style="list-style-type: none">• To let engineers see the consequences of their actions directly on the spot• To enrich our knowledge of possible threats

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Specifications: expected results

- A tool for converting an UML diagram to a general model
- A tool for performing an I-FASST analysis on a general model
- I-FASST analysis on a general model should be faster than on an XML diagram
- The possible existing research into runtime verification has been identified

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Comparison of potential general models

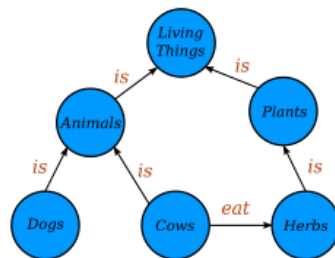
Approach	Key Characteristics	Notes / Applications
Knowledge Graphs	Graph-based structure; uses link prediction; measurable with metrics like Mean Rank, Precision, Recall	Applied in cybersecurity for threat modeling and prediction; interpretable; useful for structured knowledge representation
Vector Databases	Store data as numerical vectors for fast similarity search	Faster but may lose relationship semantics; not ideal when link fidelity is crucial
Large Language Models (LLMs)	Neural networks trained on vast textual data; excels at NLP tasks	Still less reliable than KGs for factual accuracy and long-tail entities; promising for unstructured data extraction
Other Approaches	Concept maps, conceptual graphs, semantic networks	Often based on or derived from knowledge graphs; more task-specific and less generalizable

Choosing a database

DB name	Apache2 license	Programming language	Description
ArangoDB	+	C++, JavaScript, .NET, Java, Python, Node.js, PHP, Scala, Go, Ruby, Elixir	NoSQL DB, ArangoDB Query Language, high scalability
Amazon Neptune	+	Not disclosed, used through queries	Easy to migrate data from AWS
Azure Cosmos DB	-	Not disclosed, used through queries	Multi-modal DB, Apache Gremlin language
JanusGraph	+	Java	High scalability
NebulaGraph	+	C++, Go, Java, Python	High scalability
Neo4j	+	Java, .NET, JavaScript, Python, Go, Ruby, PHP, R, Erlang/Elixir, C/C++, Clojure, Perl, Haskell	Web and desktop versions, Cypher query language, the biggest community
Microsoft SQL Server 2017	-	SQL/T-SQL, R, Python	SQL support
Oracle	-	PGQL, Java, Python	PGQL query language
OrientDB	+	Java	Both a graph DB and a NoSQL DB
TerminusDB	+	Prolog, Rust, Python, JSON-LD	Document-oriented KG
TigerGraph	-	C++	Parallel, super-fast

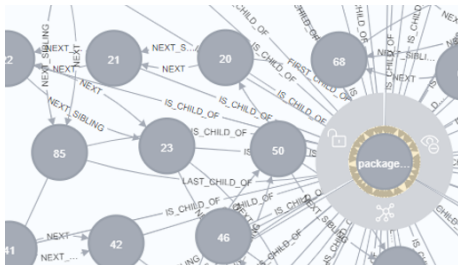
Knowledge graph

- A structure to store the data
- Has nodes and relationships between them
- Can store any type of information
- Human-readable



Choosing an approach

APOC tool



LLMGraphTransformer tool

- Shows good results
- However, it is not deterministic

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Diagram conversion

Remark

We call objects in XML file as "*entities*" with "*attributes*", and their representation in a knowledge graph as "*nodes*" with "*properties*"

- Iteratively go through the XML diagram and transform each entity and relationship
- Use a predefined set of rules

Examples

The node's label in the KG is equal to the entity's attribute "xmi:type"

Examples

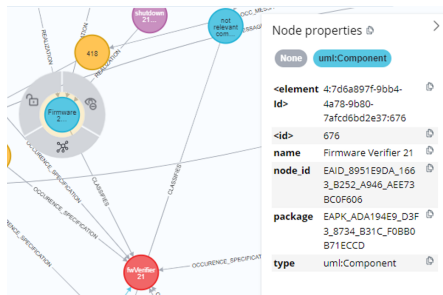
A relationship has type "OCC_MESSAGE", if it is represented in the diagram as an entity with type "uml:Message". It is an edge in the KG between two nodes of type "uml:OccurrenceSpecification"

Diagram conversion: example

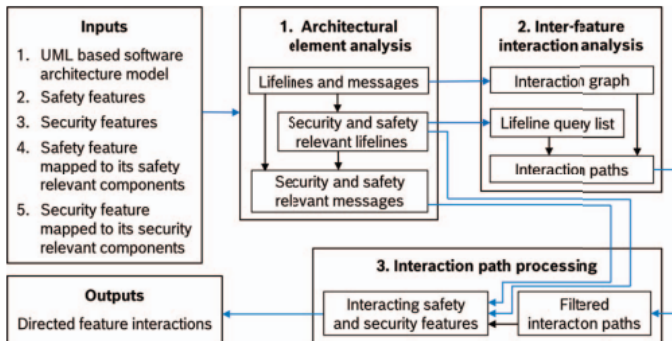
From an XML file

```
<packagedElement xmi:type="uml:Component"  
xmi:id="EAID_8951E9DA_1663_B252_A946_AEE73BC0F606"  
name="Firmware Verifier 21" visibility="public"  
isIndirectlyInstantiated="true"/>
```

To a knowledge graph



I-FASST method



Lifeline

A structural element in a sequence diagram that classifies a component. In our case, 1 feature has 1 component and 1 lifeline

1.

We say that two lifelines have the same relevance, if both of them are safety relevant, or both of them are security relevant.

2.

We say that two lifelines have different relevance, if the first lifeline is safety relevant and the second lifeline is security relevant, and vice versa.

3.

We say that a lifeline is not relevant, if it is neither safety nor security relevant.

A. Architectural element analysis — 1. Safety and security relevant activities and components

We can skip this step because we have it as input, however, it is possible to find them manually:

1. Find all nodes in the graph that have type "uml:Activity"	2. For each node, take its property "package" and define if it is a safety or security package
3. For each activity, find connected nodes and check if they have type "uml:Component"	4. Map each feature to its safety or security relevant components

A. Architectural element analysis — 2. Message exchanges

Cypher language

Cypher is a declarative graph query language that is similar to SQL in terms of graphs

Query

```
MATCH (n) - [r] → (m) WHERE r:MESSAGE RETURN n, r, m
```

(n), (m) is a template for nodes, [r] for a relationship

This query returns a list of tuples (n, r, m) where there is an edge of type MESSAGE between n and m

B. Inter-feature interaction analysis — 1. Interaction graph

Custom query	Integrated libraries	Separate graph
Can skip this step because we already have a graph	Need to integrate	Need to build
50 seconds	21 seconds	2 seconds
Custom query usage	Dijkstra's source-target shortest path	NetworkX library (the one that was used in I-FASST)

B. Inter-feature interaction analysis — 2. Lifeline query list

Here we form a list of queries (srcLifeline, dstLifeline) with different relevance

List of queries

```
(fwVerifier, fwLoader),  
(fwLoader, fwVerifier),  
(fwVerifier 2, fwLoader),  
(fwLoader, fwVerifier 2),  
...
```

B. Inter-feature interaction analysis — 3. Interaction paths

Applying queries to the graph to find paths:

Query

```
MATCH p = (n) -[*..max_length]→ (m) // to configure max path length
WHERE n.node_id = $start_value AND m.node_id = $end_value
AND ALL(r IN relationships(p) WHERE r.diagram_id =
head(relationships(p)).diagram_id) // to search in one diagram
WITH p, nodes(p) AS all_nodes
RETURN p, all_nodes
LIMIT 1 // can have multiple paths
```



C. Interaction path processing

1. Filtering interaction paths

Remark

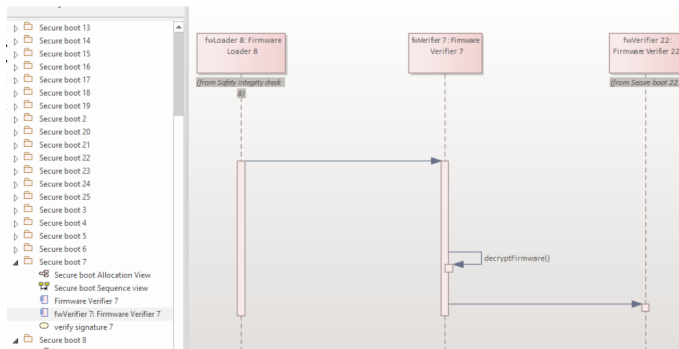
Interaction chains — interaction paths containing safety or/and security intermediate nodes

We need to avoid these paths, so we filter the set and for each path check if it has interaction chains

2. Interacting features

The rest "good" paths are returned as a function output

The unit-under-test: a diagram created in Enterprise Architect



The final version contains around 1,300 nodes and 1,400 relationships, 50 safety and 25 security features

Comparison metrics: execution time and the information obtained

I-FASST on KG

path
Secure boot 21 → Safety integrity check 44
Secure boot 10 → Safety integrity check 49
Secure boot 11 → Safety integrity check 40
Secure boot 12 → Safety integrity check 26
Secure boot 18 → NRF 6 → NRF 4 → Safety i. c. 16
Secure boot 20 → Safety integrity check 51
Secure boot 3 → NRF 2 → Safety integrity check 27
Secure boot 6 → Safety i. c. 47
Secure boot of automotive firmware → Safety i. c.
Safety integrity check 32 → Secure boot 24
Safety integrity check 15 → Secure boot 22
Safety integrity check 6 → Secure boot 23
Safety integrity check 8 → Secure boot 7
Safety integrity check 44 → Secure boot 21
Safety integrity check 50 → Secure boot 9
Secure boot 4 → Safety integrity check 2

Time: 1.87 seconds

I-FASST on XML

Feature 1	Feature 2
Secure boot 21	Safety integrity check 44
Secure boot 10	Safety integrity check 49
Secure boot 11	Safety integrity check 40
Secure boot 12	Safety integrity check 26
Secure boot 18	Safety integrity check 16
Secure boot 20	Safety integrity check 51
Secure boot 3	Safety integrity check 27
Secure boot 6	Safety integrity check 47
Secure boot 1	Safety integrity check 1
Safety integrity check 32	Secure boot 24
Safety integrity check 15	Secure boot 22
Safety integrity check 6	Secure boot 23
Safety integrity check 8	Secure boot 7
Safety integrity check 44	Secure boot 21
Safety integrity check 50	Secure boot 9

Time: 2.09 seconds

Comparison metrics: user experience

	I-FASST on KG	I-FASST on XML
Visual representation	+	-
Can be stored in a cloud	+	-
Changing in runtime	Easier	Harder
Complex queries	Using Cypher	Should be resolved manually
Runtime verification	Bigger potential because of the above	Is possible, but harder

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Definition

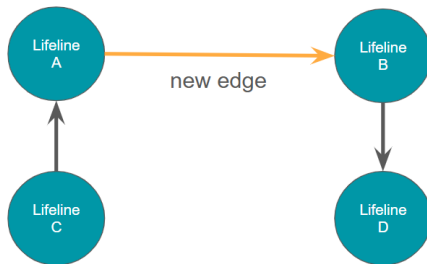
Runtime verification is a computing system analysis and execution approach based on extracting information from a running system and using it to detect and possibly react to observed behaviors (Wikipedia).

Criteria under analysis depend on current system

- E.g., out-of-bounds sensor values for safety components
- E.g., confidentiality for security components

Runtime verification, scenario 1 — a new edge is added

Case 1	Case 2	Case 3
The source and the target nodes have the same relevance	The nodes have different relevance	At least one of the nodes is not relevant: <ul style="list-style-type: none">– The source node– The target node– Both nodes



Runtime verification, scenario 1 — the same relevance

Relationship properties

MESSAGE

<element 5:7d6a897f-9bb4-4a78-9b80-7afcd6bd2e37:1155173304420533712

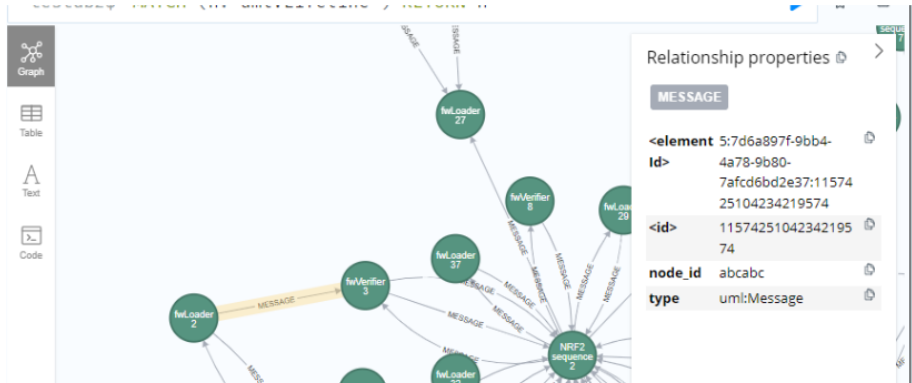
<id> 1155173304420533712

node_id new_irrelevant_id

type uml:Message

```
PS C:\Users\egorova\Documents\code\general-model-of-safsect1> .debugpy-2024.0.0-win32-x64\bundled\libs\debugpy\adapter/../../de
Edge added!
Path is not relevant, both nodes have the same relevance
```

Runtime verification, scenario 1 — different relevance

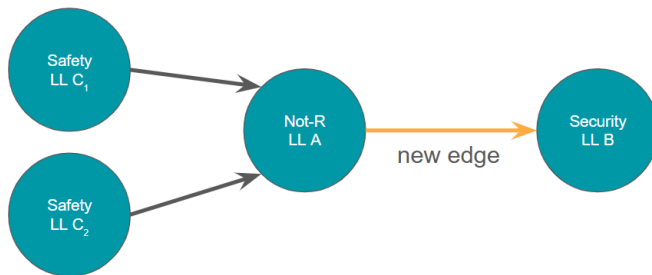


Edge added!

Path (fwLoader 2)-[:MESSAGE {node_id: 'test_edge_1', type: 'uml:Message'}]->(fwVerifier 3) is potentially relevant
PS C:\Users\egorova\Documents\code\general-model-of-safsecfi>

Runtime verification, scenario 1 — at least one lifeline is not relevant

For example, the source lifeline is not relevant, and the destination lifeline is **security relevant**



- Path 1: $C_1 \rightarrow A \rightarrow B$
- Path 2: $C_2 \rightarrow A \rightarrow B$

Runtime verification, scenario 2 — timing aspect

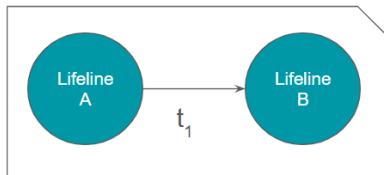


Diagram D1

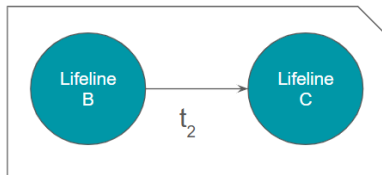
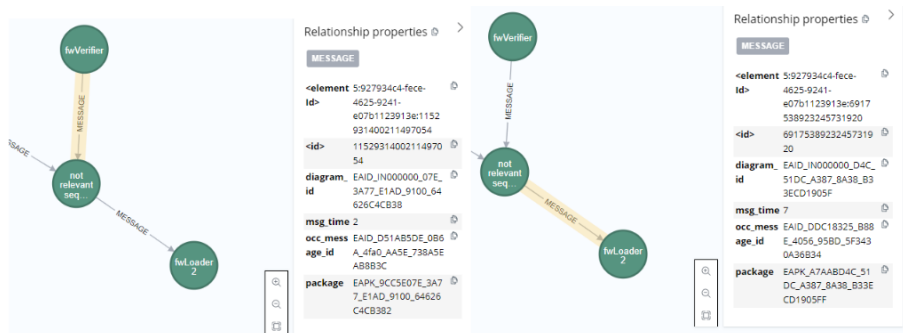


Diagram D2

- Lifeline B has to be non-relevant to avoid interaction chains
- Search for paths that begin and/or end with a non-relevant lifeline
- Try to form the paths $(A) \rightarrow (B) \rightarrow (C)$ with relation to relevance of A and C
- Timing of t_1 should be less than timing of t_2

Runtime verification, scenario 2 — example



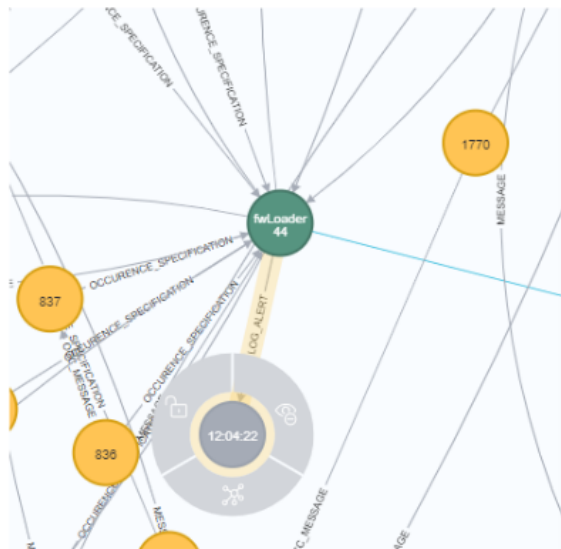
Logs aspect, scenario 1 — from KG to logs

- For example, error message from a security feature to itself
- The message that is marked as unexpected

Logs output

Alert! Anomaly Sensor value S1 is out-of-bounds detected in
EAID_LL000000_8428_2537_A50F_802085E313F8

Logs aspect, scenario 2 — from logs to KG



Node properties

LOG_INFO

<element Id> 4:7d6a897f-9bb4-4a78-9b80-7afcd6bd2e37:1

<id> 1

message Sensor value S1 is out-of-bounds

msg_time 12:04:22

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Results obtained with relation to specifications

Specification	Status	Comment
A tool for converting a UML diagram to a KG	Met	Developed
A tool for performing an I-FASST analysis on KG	Met	Developed
I-FASST analysis on KG is faster than on the XML diagram	Partially Met	The speeds are relatively equal
The possible existing research into runtime verification has been identified	Met	2 general + 2 logs scenarios

The topic still offers a vast field of research, and here are many ideas to be inspired by, for example:

- A study of the use of test case generation and its application during runtime to improve the reliability of a model
- User interface to improve the user experience
- Improvement of the I-FASST method itself
- etc...

The code is available on GitLab:



Thank you for your attention and to LCIS for this opportunity!

