

# Handling Safety and Cybersecurity Interdependency in NFV Safety Architecture With the Use of An Ontology-based Solution

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## Abstract

In case safety-critical systems face an anomaly (either intentional or not), safety and cybersecurity impact humans and environment. Thus, they affect each other and so they are considered as interdependent. An ontology-based solution for safety is needed to handle this interdependency. We propose a new safety ontology for Network Function Virtualization (NFV) framework which is able to cover reliability, availability, maintainability, and integrity-related breakdown types, since they interact and influence safety according to ENISA. Our ontology allows us to have a uniformized representation of the potential anomalies that a system and its elements can face. Based on this representation, a decision-making process takes place to avoid potential conflicts between safety and cybersecurity, in order to best handle their interdependency. The results of our implementation show that our ontology handles the safety and cybersecurity interdependency, and has little impact on decision-making time, which makes it an effective methodology for NFV framework.

**Keywords:** *Safety ontology, NFV safety architecture, Safety and cybersecurity interdependency*

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## 1. Introduction

In safety-critical systems, safety is the most significant property to be considered. This is because the main focus for these systems is to prevent harm on humans and environment. However, safety is able to interact with other properties as well. According to the ENISA standard [1] safety is a subset of the reliability, maintainability, availability, and integrity properties. In this way it is understood that safety has the ability to interact with these aforementioned properties while considering their impact on humans and environment. Furthermore nowadays, NFV applications are expanded as they are used to various types of systems. Thus, NFV can be applied in safety-critical systems. In this case safety is an important property for NFV. In [2] an NFV application is used in a safety-critical use case which proves the importance of safety in these systems. For example NFV handles services for an autonomous vehicle. In case a reliability anomaly happens in one of the NFV services and the vehicle becomes uncontrollable it can have an impact on the people that it carries,

the people in the surrounding, and the surrounding environment itself.

However, as seen in ENISA standard the properties that interact with safety, are shared with some of the properties of cybersecurity. This makes understood that safety is also able to interact with cybersecurity. Thus, the functionalities of safety are able to influence and violate the ones of cybersecurity. Likewise the functionalities of cybersecurity can affect the ones of safety. As an outcome it is possible to consider safety and cybersecurity as interdependent. As an example, in order for cybersecurity to mitigate an anomaly, it asks for a re-launch of a Virtualized Network Function (VNF). This is issued to the NFV Orchestrator (NFVO) module. This is because this module is the responsible one for implementing all the issued orders. At that moment safety understands that this action goes against its safety measures and blocks the NFVO from issuing this specific re-launch.

In order to prevent any safety and cybersecurity violations, it is needed to be able to differentiate the safety anomalies from the cybersecurity ones. An ontology-based solution is a good way to automate this process. Thus, it is possible to find ontology-based solutions for each one of the safety-related properties

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independently. However, in the literature there are no ontology-based solutions for safety considering all the properties related to it as a whole. Moreover, there are no ontology-based solutions that provide a safety and cybersecurity interdependency. This has the effect of limiting the decision-making process that is used for distinguishing the anomalies created in a system. Furthermore, this prevents from taking into consideration the interdependency of safety and cybersecurity.

Thus, it is understood that in order to ensure safety in a NFV framework, there are specific challenges to be addressed. These challenges deal with: (i) the detection and mitigation of a variety of safety anomalies in a more comprehensive way, and (ii) the management of safety and cybersecurity interdependency. In order to handle safety in a NFV framework, an orchestrator is needed which is able to detect reliability, availability, integrity, maintainability-related anomalies with respect to safety. Ontology is a good option for addressing this issue since ontology is an explicit specification of a conceptualization where the knowledge of a domain is represented in a declarative formalism [3]. This makes it possible to represent the different types of anomalies in relation to safety. According to this uniformized representation the reasoner (piece of software) is able to infer logical consequences. These consequences make it possible to understand whether a safety-related anomaly is also a cybersecurity-related one.

To this end our solution proposes: (i) a new ontology for ensuring safety in NFV framework, and (ii) specific rules to be used by the reasoner. Our proposed ontology is used by an orchestrator that handles safety in a NFV framework. This ontology includes: (i) the description of safety and the properties related to it (i.e., reliability, availability, maintainability, and integrity) as classes, (ii) the concerned elements for each property as sub-classes, and (iii) the breakdown types for the potential adversities as object properties. Our proposed rules allow us to automate the decision-making process. This is because the reasoner needs the rules to make a decision. According to this decision a NFV safety orchestrator is able to modify the plan of mitigation. With this modification it is possible to avoid potential safety and cybersecurity conflicts.

The remaining of this paper is organized as follows. Section 2 reviews the relevant works of ontologies. Section 3, introduces our proposed ontology. Section 4 provides the rules for supporting the decision-making process. Section 5 presents the evaluation of the feasibility of our proposed ontology. Section 6 provides the results and their analysis. Finally, the last section concludes the study and it discusses possible future work.

## 2. Related Work

In general, ontologies are used for system modelling, since they are capable of describing a whole system with its components and sub-systems. This is because an ontology is expressed as the study of what exists in a certain context [3].

### 2.1. Ontologies for safety-related properties

As follows it is possible to provide the related work with respect to ontologies for all the safety-related properties but also the integration of safety and cybersecurity. These ontologies are provided in general.

#### 2.1.1. Safety Ontologies

In relation to safety, ontologies are commonly used for obtaining safety risk knowledge and handling safety management. For safety risk knowledge, it is possible to develop an ontological method which organises this knowledge into seven unified classes (i.e. project, construction activity, risk factor, risk, risk grade, risk consequence, and risk prevention measure) [4]. For handling the risk management, an ontology with a case-based reasoning is used as a decision-making approach for safety risk management [5]. Moreover, safety ontologies are able to represent specifically extracted information from databases. In this way ontologies can assist for identifying additional capabilities of these information [6].

However, ontologies can be integrated with other technologies, algorithms, or methodologies in order to enhance their capabilities. For instance ontologies can be integrated with computer vision algorithms to develop knowledge graphs that can automatically and accurately recognise hazards even when they are subjected to change [7]. Another example is when ontologies can be combined with wireless networks to identify potential hazards [8].

#### 2.1.2. Reliability Ontologies

Reliability with respect to ontologies is expressed as a way to make ontologies reliable, or to use ontologies for increasing reliability in various systems. Agile methodology uses agile principles and practices for ontology development. In this way it is possible to utilize software engineering to build reliable ontologies [9]. Moreover, ontology alignment is a way to create reliable ontologies. In [10] machine learning techniques are used to automatically align ontologies to make them more reliable.

However, ontologies are able to be used in various methodologies in order to provide a variety of different types of reliability. In general an ontology-based text mining methodology is able to maximize system reliability, since it is able to extract knowledge from databases [11]. There are many technologies and methods in order to use semantic web and ontologies for providing reliable services. This is because the use of semantic technologies in the modeling of a multi-agent system, are very effective in increasing coordination and interoperability, as seen in [12]. Furthermore, ontologies are able to assist into making the numerical simulation techniques more reliable. This can happen with ontology-based text and data mining techniques, as seen in [13].

#### 2.1.3. Availability Ontologies

Ontologies for ensuring availability are not widely researched in the literature, up to our knowledge. However, in [14], ontologies are used to provide and ensure heterogeneous knowledge for a specific concept. By combining these ontologies with optimization algorithms, it is possible to provide high data availability.

To sum up, availability is closely linked to reliability and maintainability. Once a system is reliable and maintainable, then it is possible to satisfy availability [15].

#### 2.1.4. Maintainability Ontologies

Maintainability is an attribute that is included in dependability. In order to be able to understand all attributes of dependability but also to compare them, it is possible to use a dependability rating ontology [16]. Thus, it is possible to obtain knowledge about the attribute of maintainability but also in relation to the other attributes. Moreover, ontologies can be created by extracting them from other ontologies or by creating them from scratch. The approach to develop an ontology is able to affect the maintainability. Thus, the evaluation of the ontology development is very important. In [17], the authors propose a methodology for evaluating ontology development from scratch.

Furthermore, it is important to be able to create maintainable ontologies. For achieving this a methodology is proposed in [18] which is able to construct ontologies using a template-based approach for ontology modelling and instantiation. However, ontologies can be also used to enhance maintainability in a system. In [19] an ontology model is proposed to facilitate maintenance strategies selection and assessment. And in [20] ontologies are used for data accessing in order to enhance system maintainability.

#### 2.1.5. Integrity Ontologies

Ontologies can be used for ensuring integrity in a system. This can happen with a framework that is able to leverage an ontology to provide representation of semantically enriched data, as seen in [21]. It is also important to be able to evaluate the ontologies with regard to integrity. In [22] an ontology-based evaluation system is proposed which is a new ontology framework of leverage knowledge modelling. This creates an easy-to-use tool for quantitative identification for integrity by combining ontology and semantic web rule language rules.

However, ontologies need some constraints in their analysis in order to be able to focus on certain attributes. One of the ways that ontology accesses data is by querying via query translation. However, constraints in general in this way of accessing data is not represented. For this reason in [23] a framework for querying data that exploits information with regard to integrity constraints is proposed for ontology-based data access. It is also possible to extend the ontology-based data access into including integrity constraints, as seen in [24].

#### 2.1.6. Confidentiality Ontologies

Specifically ontologies dedicated to confidentiality are not widely researched in the literature, up to our knowledge. However, confidentiality can be found in the ontologies that cover all attributes of the cybersecurity approach of Confidentiality, Integrity, and Availability (CIA). In [25], an ontology is developed that targets a requirement based threat analysis. These requirements refer to the attributes of CIA, where confidentiality is included.

#### 2.1.7. Safety and Cybersecurity Ontologies

Safety and cybersecurity are two different concepts, and so their ontologies are composed of different elements and objects. In [26] it is attempted to link safety and cybersecurity objectives in an ontology in order to gain better theoretical understanding.

In order to build ontologies, it is possible to extract them from already existing ones and then expand them. In this way, safety ontologies can be expanded to include also cybersecurity. In [27] an ontology that already represents safety, is expanded to consider also cybersecurity for the early stages of a system life cycle. Like this it is able to gather and rank operational needs, assess the feasibility of the desired solution, and pinpoint any technological gaps. Moreover, in [28] a functional safety ontology is improved to consider attack scenarios. In this way an ontology-based model for functional safety and cybersecurity verification and validation is proposed.

Finally, [29] attempts to integrate safety and cybersecurity in an ontology. This is different from the previous because the previous expand an already existing ontology to consider also cybersecurity and they consider the early stages of a system's life cycle or the verification and validation process. While this safety and cybersecurity ontology that is based on formal methods, is able to represent the reaction of the system in different kind of scenarios.

### 2.2. Cyber-Physical Systems

With the use of ontologies, it is possible to understand the relationships between components whether they are cyber or physical ones. In [30], an ontology framework is able to capture the relationships between cyber and physical systems. Ontologies have a wide range of usage, since they can be used as analysis tool and a way to build knowledge hubs. For the analysis tool usage, the Technology Function Matrix is developed based on ontologies [31]. In order to build a knowledge hub, the authors in [32], use an ontology-based structure.

#### 2.2.1. Safety Properties

In relation to safety and in order to develop an ontology which considers all the properties that it interacts with as a whole, it is needed to understand how each relevant work provides partial coverage of the safety properties. Starting from maintainability, OntoProg is an ontology-based solution which is used for correct decision-making and assisting in the implementation of the Prognostics Health Management, for mechanical machines [33]. Furthermore, adding also the availability property to maintainability, an ontological structure is provided for availability as a criticality analysis which determines the maintenance strategy [34].

In [35], the three properties of reliability, availability, and maintainability, are provided. However, each one of these properties are found in a different super-concept of the solution, which means that they are not associated. Finally, the reliability and availability properties are provided through an ontological solution for detecting and preventing the failures of the system components of Cyber Physical Systems (CPS) [36]. An ontology is used with all the CPS failures described in order to assist a multi-agent architecture to detect and identify the potential failures. And in [37], an ontology is built by transforming the results of the Failure Modes, Effects, and Criticality Analysis model into a class diagram. This ontology is utilized for detecting and preventing failures. As seen from above, the only paper that is the closest to the global image of safety, is the paper that includes availability, maintainability, and reliability [35]. This is because it is the only

solution that includes three of the properties that interact with safety.

### 2.2.2. Confidentiality Property

Up to our knowledge, confidentiality ontologies for CPS are not widely researched in the literature. However, since confidentiality is a sub-property of dependability according to ENISA, it is possible to find ontologies that consider confidentiality for CPS in ontologies that concern all attributes of dependability. In [36] an ontology that concerns all attributes of dependability is used to consider various failures.

Additionally, confidentiality is also a sub-property of trustworthiness according to ENISA. Thus, it is possible to find ontologies that consider all attributes of trustworthiness. In [38] SIMON is an ontology framework that is able to ensure trustworthiness and by extension all of its attributes.

### 2.2.3. Safety and Cybersecurity Interdependency

In order to build an ontology that handles the safety and cybersecurity interdependency, it is needed to see if there are any research papers in the literature that cover this topic. However, in the literature there are no papers for safety and cybersecurity interdependency in relation to CPS. In the literature, most of the papers for trustworthiness in CPS, use the NIST CPS [39] standard and none of them is using the ENISA one. In NIST CPS safety, security, and reliability are sub-groups of trustworthiness, while cybersecurity with the CIA approach, is sub-groups of security. For example, in [40], a framework is provided for reasoning about NIST CPS trustworthiness in CPS, which combines ontology-based reasoning and answer set programming. And in [41], an ontological design and verification framework is presented, which captures the relationships between cyber and physical components in CPS. Once again, NIST CPS trustworthiness is considered.

Furthermore, there is also STRAM, which is one more framework for trustworthiness [42]. According to STRAM, security and trust are its sub-groups. Safety and reliability are sub-groups of trust, while cybersecurity is a sub-group of security. Both NIST CPS and STRAM consider all of our properties separately and do not associate them. Moreover, in line with NIST and STRAM, safety and cybersecurity share no common properties. This makes us understand that by using NIST CPS or STRAM there is no way to associate safety and cybersecurity in an interdependent way. However, ENISA gives us an image of the properties that interact with safety, as well as the properties that interact with cybersecurity. Moreover, ENISA also shows the two shared properties between safety and cybersecurity, according to which it is possible to build an architecture that provides safety and cybersecurity as interdependent.

Up to our knowledge, it is possible to distinctively find ontologies for the needed properties in relation to safety in the literature. However, there are no papers for a safety ontology which includes all the safety properties that are found in ENISA. Furthermore, it is difficult to handle safety and cybersecurity interdependency through the properties of trustworthiness that are found in ENISA. And so, a new ontology for safety is needed to handle this interdependency.

## 3. Ontology-based Solution

This section presents a new safety ontology. This ontology is used by an orchestrator that ensures safety in a NFV framework. Our proposed ontology is able to: (i) describe a variety of different breakdown types related to safety, and (ii) help the decision for the best reaction to safety-related anomalies while considering the safety and cybersecurity interdependency. Our ontology-based solution is written in Ontology Web Language (OWL). This is because it provides greater content interpretability, in comparison to eXtensible Markup Language (XML) and Resource Description Framework (RDF). OWL language facilitates the expression of knowledge and it also provides the means to reason with this knowledge.

As seen in [43], there are many advantages that ontologies bring. These advantages are: (i) the modelling clarity which refers to the clear description, (ii) the choice of specificity level which refers to the level of the detailed representation of the content, (iii) the systematicity in information retrieval which makes it possible to access classes and sub-classes to get information, (iv) the systematic and coherent definitions where the conceptual information are organized and clarified, and (v) the dynamicity as the ontology is able to represent the concept evolution through time. More specifically our safety ontology is chosen because of its capabilities to model safety and its properties in a clear way with coherent definitions. Moreover, the possibility to access classes and sub-classes in order to retrieve information supports the process of making decisions. These decisions are able to manage the safety and cybersecurity interdependency.

Our safety ontology provides the description of a representation of safety and the properties that it interacts with it according to ENISA. More specifically all of these properties have the ability to cause anomalies which affect humans and environment. For this reason the safety principle is decided to be the core of our proposed ontology.

All things considered, a NFV framework consists of a variety of modules. Different orchestrators are able to handle these modules. Thus, for ensuring and handling safety in a NFV framework, an orchestrator is needed. This orchestrator requires a way to identify whether a safety-related anomaly is also affected by cybersecurity. This is where our proposed ontology takes action since with the help of the reasoner it is able to identify whether an anomaly is both safety and cybersecurity-related. In order to reach to this outcome the reasoner uses: (i) our proposed ontology, and (ii) our safety and cybersecurity interdependency rules. NFV safety orchestrator needs this outcome in order to understand whether the mitigation plan creates violations to cybersecurity functionalities during mitigation.

### 3.1. NFV Safety Orchestrator

STELAC is a new safety orchestrator, which consists of the modules: (i) Sense, (ii) Triage, (iii) Explore, (iv) Learn, (v) Act, (vi) and Conflict manager. Figure 1 provides also the connectors that our safety orchestrator uses, to communicate with the rest of the orchestrators. Our proposed orchestrator is able to support applications that require safety and cybersecurity through the anomaly that is appeared in one of the modules of NFV framework. Thus, it is understood that we refer to any fault and unintentional anomaly, or even an anomaly that is coming from any cyber-attack.

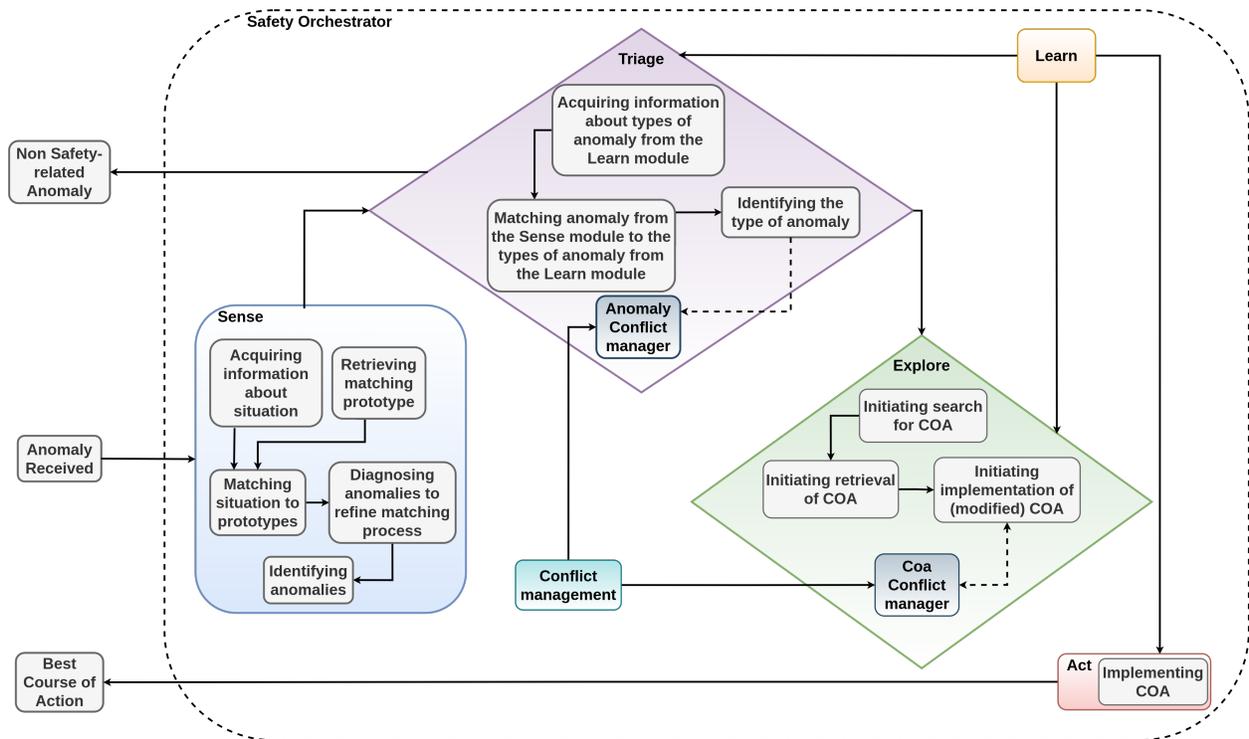


Fig. 1. New safety orchestrator for NFV framework.

The sense module is responsible for identifying the potential anomalies to understand the exact problem. More specifically our safety orchestrator receives the anomaly notification from each one of the orchestrators, through the appropriate connector in Figure 1 (i.e., anomaly received). This notification is handled by the sense module. It identifies the anomaly by a procedure of acquiring information about the situation of the component in issue while retrieving matching prototypes of anomalies. Then it performs diagnosis processes to refine the matching prototypes to the specific anomaly. And finally it is able to identify the exact anomaly.

Triage is the module that receives the identified anomaly and is able to distinguish its type. More specifically if the anomaly is not safety-related, it is sent to be handled by the appropriate orchestrator (i.e., reliability, availability, maintainability, or integrity orchestrators). Our safety orchestrator is able to send back this anomaly information through the connector that is the non safety-related anomaly as shown in Figure 1.

The explore module investigates and decides the course of action that is the most appropriate for the specific anomaly. This happens with the help of the learn module. More specifically, this module starts searching in the learn database for the most appropriate course of action according to the anomaly. The best course of action is retrieved after it has been identified. Further, it initializes the implementation of the chosen course of action. And finally it sends the decided course of action to the act module.

Up until now, the learn module is a database with information about the types of anomalies and courses of action. Furthermore the act module is responsible for implementing the best response to the specific anomaly through the other orchestrators. Finally, the course of action is implemented through the orchestrator that is affected by the anomaly. As shown in Figure 1, the course of action

is sent to the affected orchestrator to be implemented, through its designated connector (i.e., best course of action). A simple example of this course of action is the order of a VNF migration while blocking the cybersecurity orchestrator to act on the same anomaly. This blockage is happening through the availability and integrity orchestrators since they are the two common properties between safety and cybersecurity.

The conflict management module handles the safety and cybersecurity interdependency. This can happen by evaluating the decided anomaly and course of action to understand whether a possible safety and cybersecurity conflict is about to happen. This conflict refers to any potential functionality violation against cybersecurity during anomaly mitigation. More specifically conflict management is composed of the functions: (i) anomaly conflict manager, and (ii) Course Of Action (COA) conflict manager. It is the anomaly conflict manager that is equipped with the NFV safety ontology because of the need to identify whether safety-related anomalies also affect cybersecurity.

It should be mentioned that our proposed solution is also protected against false positives and true negatives. This happens once the information about a potential anomaly is sent by the reliability, availability, maintainability, integrity, or confidentiality orchestrators to the safety orchestrator. More specifically the nature of this information is identified in the sense module. In case the anomaly information is identified as a false positive or true negative then it is rejected and no further actions are performed. However, the rates of the identification of the false positives and true negatives have not been measured since they are rejected in the initial module of our orchestrator and there is no process done for them.

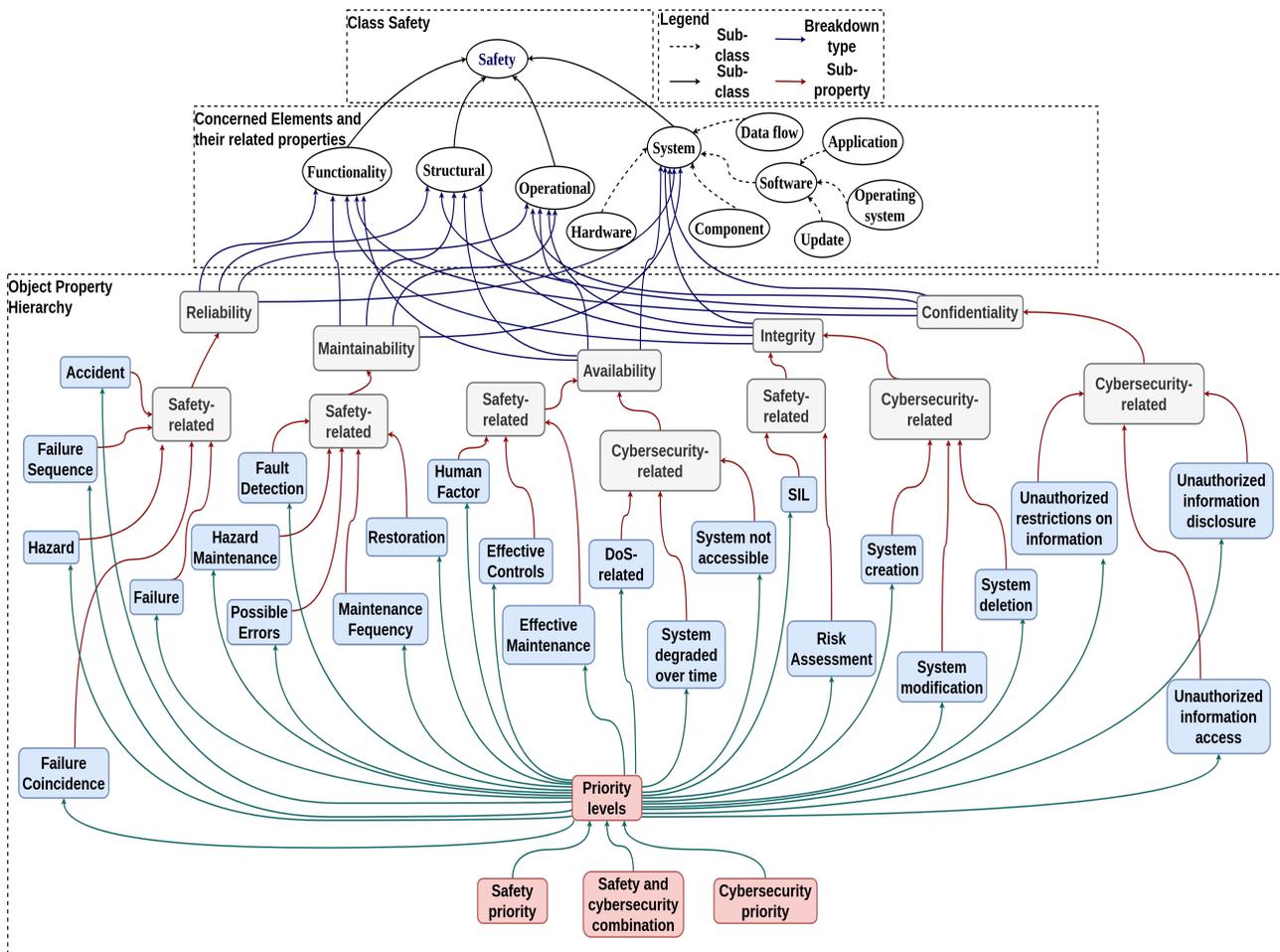


Fig. 2. Safety ontology

### 3.2. NFV Safety Ontology

In practice our ontology consists of three parts: (i) the class of safety, (ii) the concerned elements, and (iii) the object properties (see Figure 2).

The first part provides safety as the class of our ontology. The second part describes the elements that are affected by potential safety-related anomalies. These elements are the functionality, structural, operational, and system. Each concerned element is associated to a safety-related anomaly. Thus, these concerned elements are represented as sub-classes of safety. Finally, the third part provides the possible breakdown types related to each one of the safety-related properties as they are found in ENISA. Each one of the safety-related properties and confidentiality is an object property of the concerned elements. And each one of these properties is associated to its relation to safety, cybersecurity, and/or both. By extension each property is associated to their possible breakdown types. Each breakdown type is then assessed with respect to a priority level which is divided in safety, cybersecurity, and both.

It should be mentioned that our concerned elements and our breakdown types are extracted from the standards: NIST [44], MITRE [45], ISO 61508 [46]

In particular our ontology is able to provide three possible outcomes. The first outcome refers to the anomaly as only safety-related. The second outcome corresponds to an anomaly that is interdependent between safety and cybersecurity. In order to get this specific outcome it is needed to describe how the potential anomalies are related to the elements of the ontology. Thus, once a potential anomaly is identified as safety-related, it is then associated to its concerned element. According to the origin property of the anomaly (i.e., reliability, maintainability, availability, integrity, and confidentiality) it is possible to understand if it affects more than one property and if this impact is also creating cybersecurity-related breakdown-types. For example, a VNF stops working. This VNF handles the access management of the cybersecurity functions. This event is a reliability problem which affects also availability and integrity. Thus, the reliability-related anomaly is able to affect cybersecurity. In this case this anomaly is considered as an interdependent one between safety and cybersecurity.

Finally the third outcome deals with the affected breakdown-types since it is possible to assess the priority level. This priority level indicates that safety, cybersecurity, or both orchestrators can handle the anomaly.

In general our NFV safety ontology detects anomalies as messages that transfer the anomaly information. This information

is gathered by the appropriate agents, which reside in every module of the framework under examination. Thus, it is theoretically possible to use agents in safety-critical systems other than NFV frameworks. These agents are used to monitor and examine the modules or components of these safety-critical systems. This makes us understand that it is also possible to apply our solution in a variety of safety-critical systems, as long as our orchestrator is able to receive the potential anomaly in the form of a message.

Furthermore, since our solution is able to prevent any functionality violations while mitigating a potential anomaly between safety and cybersecurity, our solution has the potential to be integrated in a cybersecurity framework. This is because it is able to have both the safety and cybersecurity sides work together in harmony. However, further investigation is needed to verify that the both of these suggestions are feasible.

Additionally, our ontology-based solution is able to understand whether a potential anomaly is affecting both safety and cybersecurity in order to preserve their interdependency. This means that it is able to understand and consider both of these sides, which is something that other solutions or frameworks do not focus on. As seen in the related work section, there are not complete ontologies that provide coverage for the both of safety and cybersecurity. Nevertheless, there are two solutions that are closely relevant to our work, which are the 28 and 29. Compared to the solution in 28, our solution is able to consider all stages of a system's life and not just the early ones.

Moreover the solution in 29 is the most relevant to our work, since they also consider the reaction to different scenarios. However our solution focuses on mitigating potential anomalies and not just faulty or attack scenarios. According to our understanding one scenario may be composed from multiple anomalies. This means that the solution in 29 focuses on reacting to a situation that has already affected many different parts of the system under examination. In comparison our solution handles each one anomaly that affects any potential VNF. This is more effective because every each one anomaly that appears is treated, and so the anomalies do not propagate nor they get summed up.

#### 4. Interdependency Rules

Certain rules are defined in order to handle the safety and cybersecurity interdependency. Based on the rules the reasoner is able to make decisions for eventually avoiding potential safety and cybersecurity conflicts. There are two types of rules. The first type is composed of three statements concerning: (i) the type of breakdown, (ii) the relation of the anomaly to safety, cybersecurity, or both, (iii) and the affected object property (reliability, availability, maintainability, confidentiality, and integrity-related). According to these rules it is possible to identify whether cybersecurity is affected through availability and integrity, but also to see how cybersecurity impacts safety through all the safety-related properties. And the second type of rules is composed of two statements considering: (i) the outcome of the first rules, and (ii) the breakdown type. Thus, it is both of these types of rules that are used to automate the process of inferring, during decision-making.

Safety and cybersecurity interdependency rules are the ones that the reasoner uses to understanding whether a safety-related anomaly is also affecting cybersecurity and vice versa. But also

these rules are used to get the indication of which orchestrator between safety and cybersecurity is prioritized for mitigating the anomaly.

##### 4.1. Safety and Cybersecurity Rules

In general in order to create these rules it is taken into account the type of the breakdown and its impact to cybersecurity. More specifically the type of the breakdown is associated to: (i) the element that is affected by the anomaly, and (ii) the specific breakdown type. The impact to cybersecurity refers to the object property that is affected by the specific anomaly. Thus, it is understood that there are three important terms. These terms are: (i) the fact that the safety-related anomaly is coming from a cyber-attack (*C.A.* from cyber-attack) or has the same effect, (ii) the sub-property of the specific object property (*B.T.* from breakdown type), and (iii) the object property that is also affected by the anomaly (*O.P.* from object property). It should be mentioned that our proposed rules are considered only when the event is coming from a cyber-attack or has the same effect. Hence, each rule (*I.R.* from interdependency rule) is a set of three statements referring to these terms, with the form as seen in (1).

$$I.R. = C.A. + B.T. + O.P. \quad (1)$$

As an example, a VNF at a production unit handles the working time scheduling between humans and robots. This VNF is cyber-attacked and stops working. This is a cybersecurity anomaly. However, it also impacts safety since humans may be harmed. Thus, it is a safety-related anomaly which comes from the reliability property. This anomaly affects the functionality concerned element and it can cause the accident breakdown type. It also impacts the availability, confidentiality, and integrity object properties. Thus, in this case the corresponding rule is seen in (2).

$$I.R. = C.A. + Accident + All \quad (2)$$

All is seen in (3).

$$All = Integrity + Availability + Confidentiality \quad (3)$$

All things considered, it is understood that each safety-related property has a total number of safety and cybersecurity interdependency rules. To calculate this total number, it is needed to calculate first the total number of our proposed rules for each one of the safety-related properties. In order to make this calculations we created the formula (4).

$$I.R._{tot_x} = C.A. \times \sum_{B.T.} \times \sum_{O.P.} \quad (4)$$

In formula (4): (i)  $I.R._{tot_x}$  corresponds the total number of the interdependency rules for each one of the safety-related properties, (ii)  $x$  is substituted by the *re* for reliability, *ma* for maintainability, *in* for integrity, *conf* for confidentiality, and *av* for availability, (iii)  $\sum_{B.T.}$  corresponds to the sum of the sub-properties for each one of the safety-related properties, (iv)  $\sum_{O.P.}$  refers to the sum of the possible object properties affected for the specific anomaly, and (v) *C.A.* is equal to one since it is the Boolean true. It should be mentioned that reliability and maintainability are able to impact cybersecurity. However, availability and integrity are able to affect safety.

For each of the reliability and maintainability the  $\sum_{O.P.}$  is equal to four. This is because these properties can have 4 different possibilities of impacting cybersecurity. These four different possibilities are through availability, integrity, confidentiality, or all. For integrity it is possible to impact safety through availability, reliability, or maintainability. Thus, the  $\sum_{O.P.}$  is also equal to three. For availability it is possible to impact safety through integrity, reliability, or maintainability. Hence, the  $\sum_{O.P.}$  is also equal to three. Finally for confidentiality it is possible to affect safety through reliability, maintainability, availability, and integrity. Thus, the  $\sum_{O.P.}$  is equal to four.

Thus, the calculated total number of interdependency rules for: (i) reliability is seen in (5)

$$I.R_{tot\_re} = 1 \times 5 \times 4 = 20 \quad (5)$$

(ii) maintainability is seen in (6)

$$I.R_{tot\_ma} = 1 \times 5 \times 4 = 20 \quad (6)$$

(iii) availability is seen in (7)

$$I.R_{tot\_av} = 1 \times 6 \times 3 = 18 \quad (7)$$

(iv) integrity is seen in (8)

$$I.R_{tot\_in} = 1 \times 5 \times 3 = 15 \quad (8)$$

(v) confidentiality is seen in (9)

$$I.R_{tot\_conf} = 1 \times 3 \times 4 = 12 \quad (9)$$

The total number of rules to manage the interdependency between safety and cybersecurity is calculated in the following formula. In this formula the total number of the interdependency rules is the sum of each one of the interdependency rules of the safety-related properties. Thus the total number of the interdependency rules is seen in (10).

$$I.R_{tot} = I.R_{tot\_re} + I.R_{tot\_in} + I.R_{tot\_av} + I.R_{tot\_ma} + I.R_{tot\_conf} = 85 \quad (10)$$

## 4.2. Priority Level Rules

This type of rules depends on the outcome of the previous reasoning and rules since it is taken into account the type of the anomaly. Thus, there is only one term for this rule which refers to the related types of the anomaly ( $R.T.$  from related-type). Each rule ( $P.L.$  from priority level) is equal to this one term as seen in (11).

$$P.L = R.T. \quad (11)$$

In case the potential anomaly is safety-related and does not affect cybersecurity then this rule results that the safety orchestrator has to mitigate this anomaly. In the second case where the safety-related anomaly is impacting cybersecurity, then the outcome of the priority level rule is that both safety and cybersecurity need to handle the anomaly. In the third case where an anomaly is confidentiality and it also affects safety, then the priority level rule decides that only the cybersecurity orchestrator is to handle this anomaly.

The total number of this rule is equal to the sum of the possible related types for an anomaly. The related types for an anomaly are: (i) safety-related, (ii) cybersecurity-related, and (iii) safety and cybersecurity-related. Thus the total number of the priority level rules is seen in (12).

$$P.L_{tot} = \sum_{R.T.} = 3 \quad (12)$$

An example is provided for better understanding. A VNF handles the emergency protection of a safety-critical system. This kind of protection refers to shutting down the functionalities of a robotic arm that works along with people. This VNF suddenly stops working, which means that there is no human protection against the robotic arm. This is identified as a reliability anomaly which is related to safety. This is because the VNF fails to operate.

However, in the specifications of our system this specific VNF is able to cause our system to degrade over time, in case it stops working. The breakdown type of system degraded over time is a cybersecurity-related one. According to our ontology this is a breakdown type that is sub-property of availability. This is because the VNF is not available anymore. Thus, it is understood that a safety anomaly has the same effect as a cyber-attack to our system and that both safety and cybersecurity are affected. The reason is that the common orchestrators between the safety and cybersecurity are both triggered to detect an anomaly, while asking from the both of the safety and cybersecurity orchestrators to act on the issue. As a result, a reliability anomaly has impacted an availability one.

Thus, based on the priority level rules, our reasoner decides that the priority level outcome is for both safety and cybersecurity orchestrators to act upon the anomaly. More specifically, it is suggested that the safety orchestrator can handle the safety-related anomaly, while the cybersecurity orchestrator can handle the cybersecurity-related anomaly.

Considering everything, in Figure 3 it is possible to see the functioning model of our solution. Our model consists of three parts which are the ontology, the reasoner, and the outcome. Our ontology is represented in OWL in order to fully describe the whole knowledge of safety in one common language. Each property of safety and confidentiality corresponds to a specific set of rules. The reasoner is able to make decisions based on these sets of rules and to provide an outcome that best handles the interdependency between safety and cybersecurity.

## 5. Evaluation and Results

For the implementation and evaluation phase a testbed is constructed with the intention to test our proposed safety orchestrator. The parameters that are used for implementing and running our testbed are provided in Table 1.

Open Source MANO (OSM) handles all the VNF by using the VNF and NS descriptors for instantiation. And Openstack handles the whole architecture of the servers. Furthermore, each orchestrator of our proposed safety architecture corresponds to one VNF and one instantiated VM. It is possible to access these VM via openstack.

Our use case is provided in Figure 4. Free5GC includes the functions: (i) Network Repository Function (NRF): serves as a

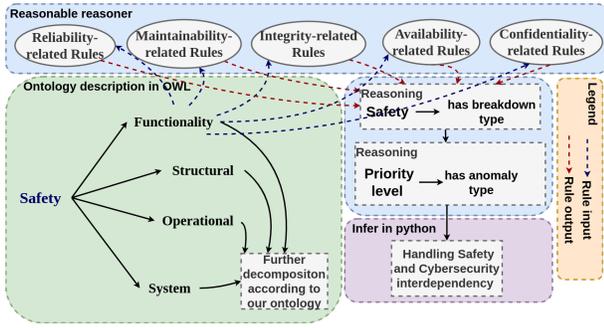


Fig. 3. Handling safety and cybersecurity interdependency, with the combination of our ontology and a rule-based reasoner

The parameters of our testbed	
Virtual Machines (VM)	6
Processor	Intel core i7
Clock rate	4.6GHz
Threads number	16
RAM	15.744 GB
Swap memory	15.6 GB
OS	Pop OS 20.04 focal (Linux)
RAM for each VM	6 GB
vCPU for each VM	3

Table 1. All the parameters that are used for testing our testbed.

central repository for virtualized functions, (ii) Authentication Server Function (AUSF): supports the authentication of an entity that attempts to access a network, (iii) Access and mobility Management Function (AMF): manages the reachability, registration, mobility, and connection, (iv) Session Management Function (SMF): controls the session, and (v) User Plane Function (UPF): serves for the part of the network that carries the data traffic.

In our testbed the Free5GC core corresponds to the first server and it represents the various VNF of a NFV framework. Each one of these functions corresponds to a VNF. Each one of these VNF is able to generate anomalies which are related to virtualized function and service issues with respect to NFV framework. For example, the VNF which corresponds to UPF function is not able to migrate. In this way it is possible to simulate the issues of a NFV with respect to VNF. This is structured in a docker environment with each VNF occupying a container which uses Ubuntu 20.04.

For this to happen, it is possible to create a database of known anomalies related to each one of these functions. Then, the VM that includes them is able to choose one of the anomalies of the database. This happens with a python script that chooses one of the anomalies every 100 milliseconds. This is because we want to: (i) be able to send to the safety orchestrator a variety of anomalies to understand how they are being treated, and (ii) handle many anomalies in a very short time to understand whether our solution is able to treat all of these anomalies successfully.

For performing a test closer to reality, a fault injection tool is used to create anomalies in each VNF of our testbed. This means that instead of having each VNF generating an anomaly by itself, now this tool is able to insert the VNF and fabricate an anomaly. For achieving this, a saboteur is used which is a python tool for injecting a variety of anomalies in the many VNF of our testbed. These anomalies can vary from internal faults that are associated

to each orchestrator to even network faults in the communication between the many VNF of our testbed.

Once the fault is injected a monitoring agent is able to identify this specific anomaly. This agent is based on the psutil which is a python tool that is able to monitor the python functions that run in a machine. This agent is able to see that a specific function has stopped running, in order to identify its associated anomaly and send it to the appropriate orchestrator.

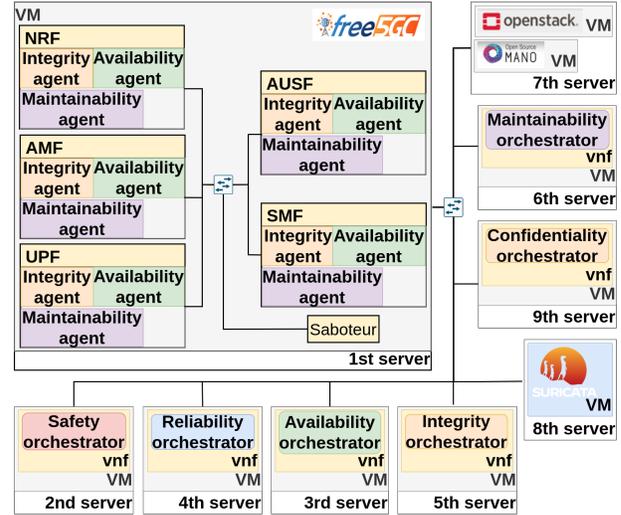


Fig. 4. NFV Safety architecture use case

Furthermore, the safety orchestrator needs to be able to receive the anomaly messages from the rest of the orchestrators. For this to happen a client-server architecture is implemented. This architecture uses the WSGIserver technology. According to this architecture: (i) safety orchestrator is the server which receives the anomaly message, treats it, and sends back the best response, (ii) and each one of the rest of the orchestrators is the client which sends the anomaly message, and then receives the reply.

Furthermore, the server sends two types of messages to the client: (i) one message is for sending the course of action to the act module of the client in order to implement them in case that the anomaly type is safety-related, (ii) and the second message is for sending back the anomaly information in case that the anomaly type is not safety-related. In case of a safety-related anomaly it is possible to handle the interdependency between safety and cybersecurity. This is because the decided course of action is already verified from the reasoner feature in the conflict management module of the safety orchestrator.

The anomaly messages are pulled by a REST API, as it retrieves the message with a GET request. Each anomaly is therefore addressed in the URL, which makes the safety orchestrator able to access and process the message. In this way the objects are retrieved by this specific URL. At first, they are treated to find the appropriate course of action. And then, they are sent to the Reasonable reasoner in the conflict management module for handling the safety and cybersecurity interdependency.

In our case our implementation handles the anomalies in parallel. As our orchestrator is based on python, gevent library is used to provide several processes in parallel. In this way each

one of the anomaly messages is handled by the available process. In case all the processes are running with an anomaly message, then the next message waits until one of the processes is available again. For our implementation we used 5 processes, since this is the indicated number of them that is provided by gevent.

In order for safety to be achieved, it is needed to analyse the results obtained from our implementation. These results are the type of the anomaly and the decided course of action that ensure safety and cybersecurity interdependency. Thus, safety is achieved once the anomaly message is treated and the course of action is decided and sent back to one of the rest of the orchestrators. Moreover, the total number of messages treated shows how effective and stable our proposal is. For this reason it is important to acquire the time that each message takes to be treated. The shorter the time that an anomaly takes to be treated, the greater the number of the messages are treated.

Consequently, the outcomes of our implementation are provided in terms of time which are: (i) the reasoner processing time: the time that the conflict management module takes to decide how to best handle the safety and cybersecurity interdependency, and (ii) the total response time: the amount of time that it takes for one of the orchestrators to receive the response from the safety orchestrator.

For realizing Figure 5, the number of the repeated tests is greater than thirty with one minute as a running time per each test. The number of anomalies treated is affected by the time that the reasoner needs to execute the rules in the safety architecture ontology, since it needs to be able to make decisions about the safety and cybersecurity interdependency. Figure 5 provides our two metrics. The total response time is in blue and the reasoner processing time is in red. Each point of the lines corresponds to the mean value of one test. For each mean value point the above and below standard deviation bars are provided.

Some points have greater standard deviation values than others, since standard deviation is affected by the number of the samples and the mean value which both change from test to test. Thus, the points with the higher mean values are the ones with the greatest standard deviation. Finally, our study provides 95% of assurance that an anomaly is treated with confidence interval bounds of 0.0072s to 0.0088s for the total response time, and 0.0018s to 0.0020s for the reasoner processing time. Consequently, the mean time values for both of our metrics, are quite low and our solution is considered stable. This makes it seem possible to use our solution for real-time systems.

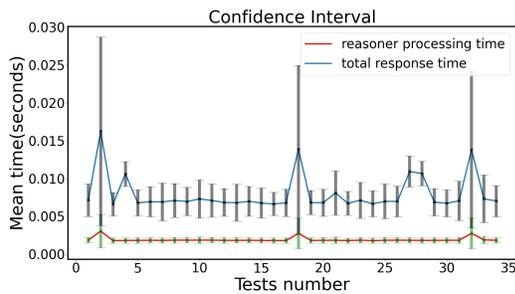


Fig. 5. Reasoner processing time and total response time

However, virtualization technologies have the ability to minimize the delay of the anomaly mitigation process.

Nevertheless they are not widely used in real applications. This means that it may take from a day to even several days to diminish the potential anomaly. This information is acquired after interviewing engineers that work on the anomaly mitigation process for real rolling stocks.

Our solution though is able to provide an anomaly mitigation in milliseconds while using virtualization technologies. Furthermore, in Figure 6 it is possible to provide the results of our two metrics that are obtained during the fault injection test. This test is repeated 65 times and once again the total response time follows the fluctuations of the reasoner processing time.

This shows that our solution is stable. Moreover the mean value of the reasoner processing time is 0.0040s, while the same value of the total response time is 0.0077s. These values are close to the confidence interval bounds that are calculated previously, and this proves that our solution is stable as it treats even only one anomaly each time successfully with a very little latency.

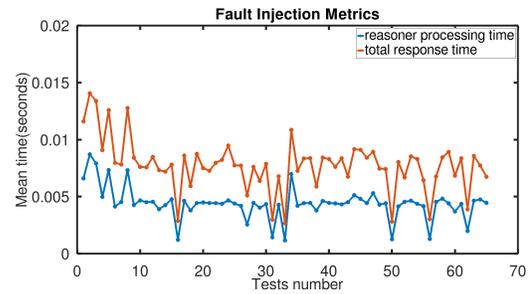


Fig. 6. Results of our two metrics under fault injection

Figure 7 illustrates the total response time. In this test the number of anomalies that are generated per second are iterated by 100 each time. It is understood that as the load of anomalies increases, the latency of the response time gradually rises. And it is almost linear. This is possible because the anomalies may be generated in one second, but the conflict management needs time to treat them all. Moreover, the anomalies are handled in parallel between five processes. However, the latency grows gently which seems to suggest that it is possible to meet the real-time constraints of many applications. Overall, this graph shows that our solution is scalable, and that it can be used for larger architectures.

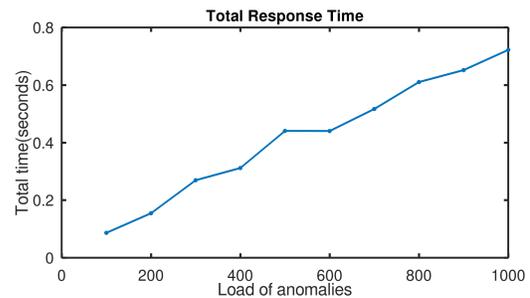


Fig. 7. Total response time's mean value in relation to different load of anomalies per second

## 6. Conclusion

Safety and cybersecurity are able to impact each other in a NFV framework, and so both of them need to be taken into consideration. Thus, it is important to be able to manage safety in a more comprehensive way. But, it is also important to handle the safety and cybersecurity interdependency. In this paper, an ontological-based solution for handling safety is proposed. Moreover, the safety and cybersecurity interdependency rules are proposed. More specifically, our proposed ontology is able to describe safety through the safety-related properties found in ENISA (i.e., reliability, availability, maintainability, and integrity). Together the ontology and the rules are used by an orchestrator that manages the safety of a NFV framework. This is because the safety orchestrator needs to understand whether a safety-related anomaly is also affecting cybersecurity. This specific information is able to help the safety orchestrator to modify the plan of mitigation, in order to avoid any functionality violations between safety and cybersecurity. In order to evaluate and test our solution, a testbed is created. This testbed is a safety and security management in 5G core network. According to the obtained results, our solution is able to ensure safety. Moreover, our solution is scalable, and it can be used in other applications.

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