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Uncertainty Dimensions of Risks in Secure and Dependable Domain

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Abstract

Most of the critical aspects for secure and dependable systems, such as safety, integrity, availability, are related to uncertainty. Literature proposes many approaches to deal with uncertainty, mainly in the area of risk management and safety&reliability engineering. However, what is still missing is a clear understanding of the nature of uncertainty that very often has produced mistreatments in the design. In this paper, we propose a conceptual model for uncertainty that can be used to deal with systems' qualities such as security and dependability. Particularly, we will consider the relation between uncertainty-risk and how risk affects several quality attributes of the system. This understanding is necessary for the evaluation of design alternatives. We use a case study in Air Traffic Management to illustrate our approach.

1 Introduction

Software systems are assuming more and more a critical role in our daily life and this introduces the need for software developers to deal more deeply with problems related to security and dependability (S&D) issues. Several approaches have been proposed in literature to develop secure and dependable systems [1, 2, 3, 4, 5] and almost all of them focus on how preserve their properties in all possible situations. Avizienis et al. [1] proposed a taxonomy to clarify the basic concepts of security and dependability that can be considered complementary to what proposed in [2, 6]. Both, however, consider risk as an uncertain event that produces a negative impact into the system [7] and that may obstruct its normal behaviour. Consequently, risks have to be identified as soon as possible in the software development process so to introduce treatments directly as part of the design.

Techniques based on ontologies and taxonomies, such as [4] and [3], have been proposed in literature to identify risk, but unfortunately they have not produced significant results for designing of secure and dependable systems. The main problem here is about how to assess *likelihood* and *severity* of risk and what kind of treatment has to be introduced to make risk acceptable for the overall design. In the dependability research community, Fault Tree Model (FTA) [5] is used to structure how a failure occurs and finally to assess the probability of the failure occurrence. FTA uses *probability theory* to compute the probability of failure. Similarly, in the security area the attack graph [8] is proposed to model how an attack is conducted and following Bayes Network rules

what is its probability. All these approaches analyse and assess uncertain events (i.e., risks, attacks, faults, errors) in terms of the likelihood. The common idea is to ensure the system is “good enough” in preserving its security and dependability under uncertainty conditions. However, there is not a clear understanding about which uncertain events one should consider. We can consider, for example, different events like

- a controller fails after 10 days of operation and this happens in 90% of cases;
- 30% of software code introduces vulnerabilities;
- a controller is hot.

All these events can be considered as risks for a system, but each of them has a different nature of uncertainty, and consequently introduce specific needs in terms of treatment. The first one refers to the variability of the controller [9], the second to the lack of knowledge [10] in detecting the vulnerability of code, and the third one to an imprecise definition of “being hot”.

In this paper, we propose a conceptual framework to model and analyse risk within the context of security and dependability. We clarify about different natures of uncertainty (i.e., aleatory, epistemic, and fuzziness/impreciseness) and how they influence properties of security and dependability. The models, we propose, can be used by an analyst to identify risks and evaluate their impact over the system according to their nature. Since security and dependability are aggregation of different quality attributes (e.g., availability, confidentiality, integrity, etc.), in our approach the risk assessment must be conducted from different dimensions.

The remaining of the paper is structured as follows. Initially, we present our understanding about basic concepts of security and dependability. We identify several types of uncertainties (Section 3). We propose a generic framework for risk assessment that is able to accommodate all types of uncertainty. (Section 4). Finally, we discuss our proposal and draw some remarks (Section 5).

2 Security and Dependability: Basic Concepts

In this section, we clarify our understanding about security and dependability as quality attributes. Starting from two US-DoD standards: Orange Book [11] and Failure-Modes and Criticality Analysis (FMECA) [12] that are considered referential works for S&D engineering community.

2.1 Quality Attributes

In Fig. 1, we propose a taxonomy for security and dependability as quality attributes. It is mainly based on the work of Avizienis et al. [1] and extended with other approaches proposed in [2, 13, 6, 14].

Information security [15] is refined into confidentiality, integrity, availability, and according to the ISO/IEC 1335 [16] we consider also authenticity and accountability. Dependability is defined as an aggregation of availability, reliability, integrity, safety, and maintainability [1]. The definition of these concepts are defined as follows:

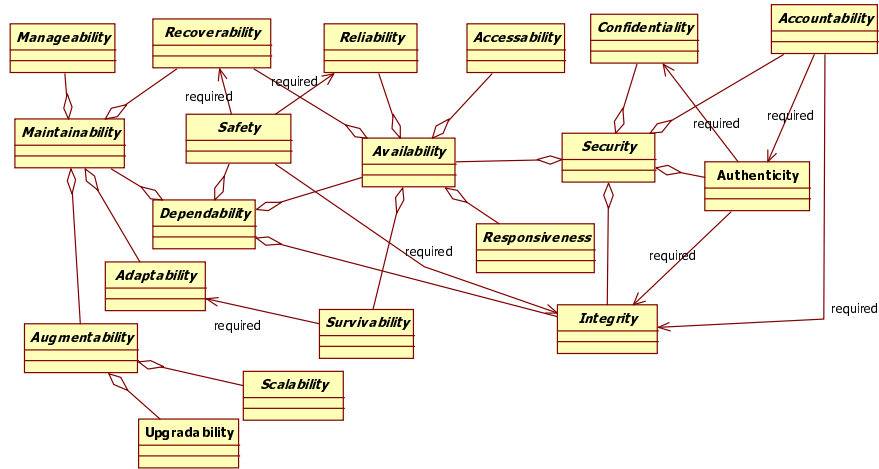


Figure 1: Security and Dependability Properties

- *Availability* is a readiness of a system (or a component) to perform its functionalities. This attribute can be achieved by five ways:
 - High *reliability* - continuity of providing correct service for a specific period;
 - Short *recovery* - repairing the system from a failure and restoring to the “good” condition before the failure occurs;
 - High *survivability* - delivering the functionalities under attacks/failures;
 - High *accessibility* - accessible by users in delivering the functionalities;
 - High *responsiveness* - users receive the response in timely manner.

Responsiveness and *accessibility* attributes are more to users’ perception and not solely depending on the systems [17].

- *Confidentiality* - the information can only be accessed by authorized agents;
- *Integrity* - the absence of unauthorized/improper system alterations. It also subsumes the absence of data modification without authorization;
- *Authenticity* - ensuring the validity of a subject identity;
- *Accountability* - ability of a system tracing back any actions;
- *Maintainability* - features that allows us to change the system after it is delivered and used. As in [1, 6], maintainability is refined into more fine-grained qualities:
 - *Manageability* represents the easiness in managing the system;
 - *Recoverability* (i.e., the same as the one at the availability);

- *Augmentability* is the easiness to upgrade the system due to the obsolete or scalable problem;
- *Adaptability* is the easiness to adjust the system according to the environment situation
- *Safety* depicts the absence of catastrophic consequence to users/environment.

Notice that these qualities are inter-related each other. A “required” relation indicates that to realizing a property (e.g., safety) the system requires another property (e.g., reliability) to be guaranteed also. Here, we note only minimal required-relations, because there could be a circumstance where safety requires other properties than the one depicted in Fig. 1. Composition relations (diamond arrows) depicted that a property, actually, is a composition from several other properties. For instance, to realise a safe system, analysts should design the system, at least, having highly reliable system, short recovery time, and making the system able to preserve its integrity. For having an accountable system, the system is required to have authentication mechanisms and to be able preserving the integrity of audit trails.

S&D attributes are related also to other quality attributes of the systems (e.g., usability, performance). For example, the security of a system usually make the system less usable by the users. However, in this situation analysts, often, capture the relation among attributes using causal-effect [18] or contribution relation [19] which is conceptually inappropriate. Indeed, there is a relationship between safety and limited-usability but it is only correlation, and not contribution nor causal-effect one. It is because the “cause” of having limited-usability is depending on the security mechanism, that is chosen, and not the state of being secure. In other word, there could be a mechanism which does not limit the system usability. For example, a system should sustain against Denial-of-Service (DoS) attacks (security). Applying a firewall that result in limiting users’ access is surely will secure the system and also limiting the usability. Conversely, having Intrusion Detection System (IDS) is able to mitigate DoS attacks without limiting users’ access.

2.2 “The Attributes” of S&D systems

As emerging from literature, reliability, confidentiality, integrity, and recoverability are the “most” important S&D properties.¹ In the following, we demonstrate how-to other S&D quality attributes can be reduced to these four properties.

Example 1. To maintain the availability of an ATM system, analysts should ensure that the system has high reliability and require short-period for recovering from failures.

Though we do not assess the accessibility and responsiveness, we may still argue about the availability of the system. However, the availability is often measured by the percentage of operated-time over a period-time. In this way, a *reliable* system may result in a low frequency of downtime, and a *recoverable* system allows us to have a short-period for each downtime.

¹In this work, we mainly focus on the first three quality attributes

Example 2. To ensure the safety of an aircraft, controllers require the sustained correct assistance (i.e., reliable and recoverable) of an ATM system, and only authorized personnels are allowed to access the system (i.e., integrity)

Notice, other aspects of an available system (i.e., reachability, responsibility, or survivability) are not sufficient to guarantee the safety operation in ATM. For instance, a high survivability system means the system keeps running though an attack/an error occurs. Actually, it is dangerous for the safety because the system usually runs in a degraded mode where the result is not 100% correct, and can give mislead information to the controllers.

Example 3. For authentication purposes, each controller has different ratings that indicates their experiences. A Novice controller should not be allowed to manage a highly-dense airspace (e.g., approach/TRACON sectors). For this end, an ATM system should authenticate the users before they use the system.

This setting requires the system to be provided with an access control (i.e., authentication mechanisms and authorization managements). In [13], the authors mentioned that there are three means to authenticate users: 1) by what users know - password, 2) by what users have - card, and 3) by what users are - biometric. Besides ensuring the *reliability* of the access control, to guarantee the authentication the system should preserve the *confidentiality* of authentication means (e.g., passwords, cards) and preserve their *integrity* against illegal modification. When the integrity of authentication means is compromised, the authenticity of users are doubtful though the access control still runs correctly.

Example 4. Controllers' actions should be accountable, especially, during an aircraft action. An ATM system always produces audit-trails for every actions of controllers and conversations between controllers and pilots. These trails is useful to perform an investigation for any incidents/accidents.

To guarantee this property, the system should provide: 1) an *authentication* mechanism to ensure the authenticity of controllers and 2) maintain the *integrity* of the audit trails from illegal modifications or fabrications [13].

Finally, these reductions, into four main qualities, are useful to reduce the number of assessments required by a system. However, it may overlook on capturing particular attacks or incidents (e.g., DoS attack). In that attack, the system is still reliable and consequently does not require any recovery but it is not available to users. For maintainability issues, we believe it is hardly possible to reduce them into reliability, confidentiality, integrity, and recoverability properties. Ideally, the maintainability should be inherent as main principles in developing the system that should be keep in mind by the developers (i.e., analysts, designers).

3 The Nature of Uncertainty

In standards like [7, 20], risk is defined as the combination of the probability (uncertainty) of an event and its (negative) consequences. In other words, risks have two

properties: uncertainty and severity. An event, which is certain or has no negative impacts to the system, should not be considered as risk, but more as problem.

... Uncertainties appear everywhere. ... When using a mathematical model careful attention must be given to uncertainties in the model. - R. P. Feymann²

Besides assessing events' impacts, it is essential to know the uncertainty associated to an event and what is its nature. Several works have been proposed about the nature of uncertainty [21, 22, 23, 24]. In [23], the authors argue that essentially uncertainty are divided into two classes: aleatory and epistemic uncertainty. The former is induced by randomness, and the latter is due to incomplete knowledge. In another paper, Smithson [22], argues that uncertainties is introduced because of vagueness. Uncertainty is resulted from ambiguity, vagueness (or impreciseness), and probability. In this paper, we consider uncertainties into three classes: aleatory, epistemic, and imprecise.

Uncertainty is always present though one perceives having a complete knowledge. For instance, we know all about a dice including possible outcomes, but still we cannot decide for certain the outcome of rolling a dice. In this example, we can categorise the event of rolling dice is exposed to an aleatory uncertainty induced from the nature of **aleatory** (i.e., variability and randomness).

Example 5. The radar engineers have, almost, full knowledge how a radar work, but they still put another radar as a backup because the main radar might still fail.

This class of uncertainty is heavily studied in the dependability community. Typically, the uncertainty arises due to spatial variation (e.g., the development site is different with the runtime site), temporal fluctuation, and development variability. Since this uncertainty may result in failures, analysts should investigate all possible uncertain events in this class. The Probabilistic Risk Analysis (PRA) [9] or fault tree analysis (FTA) [5] are useful in assessing the level of uncertainty. However, they cannot reduce the uncertainty. To deal with such uncertainties, redundancy techniques [25] (e.g., primary-secondary, main-backup, multi-version programming) are used to deal with these events.

Epistemic uncertainty is caused by incomplete knowledge.

Example 6. Aircraft hand-off procedures to adjacent sectors might fail though the ATM system operates normally. It is because analysts do not have complete knowledge about all possible actions that a controller will do for handing-off an aircraft.

Most of security problems are arisen because of the incompleteness. Security analysts are hardly possible aware of all possible threats that attackers might launch to the system. Essentially, there are two subtypes of epistemic uncertainty: 1) visible incompleteness where *we are aware about what we do not know*, and 2) blind incompleteness where *we are not aware about what we don not know*. The example 6 is one example of visible incompleteness. Essentially, it can be reduced by empirical efforts, such as: interviews, collect more data, expert judgements, etc.

²in Appendix F - Report of the Presidential Commission on the Space Shuttle Challenger Accident (In compliance with Executive Order 12546 of February 3, 1986), NASA, 1986

Example 7. In smart-card technology, attackers can extract secret keys and consequently compromise the integrity of smart-card by analysing the power consumption of a smart-card [26].

In this setting, analysts never had got an idea how such possibilities existed. Essentially, there is no such mechanism to reduce this subtype of uncertainty. Having a system with high augmentability is beneficial for this uncertainty because it should allow us to easily upgrade the system as soon as a vulnerability is discovered. Both types of uncertainty are also introducing risks to the system, but they require different treatments to reduce/to deal with the nature of their uncertainty.

The last class is **imprecise** uncertainty which is introduced due to imprecision. The imprecision can be caused by un-clear definition (e.g., ambiguity, vagueness) or the limitation of measurement/assessment techniques. In other words, this uncertainty may expose an object together previous uncertainties.

Example 8. Analysts should be able to assess the severity of the hand-off failure.

How analysts can come up with the precise value? It is common if the result is just a vague figure. In other situations, the fuzziness is introduced due to the limitation of the techniques.

Example 9. To reduce uncertainties of the event in example 6, ones may conduct interviews to some controllers (defined by a statistical sampling technique).

However, the outcome of the interviews still contains the uncertainty (called error) due to the limitation of the techniques. Essentially, the imprecise uncertainty is not the source of risks, but they play a role in defining the extent of risks. In other words, the fuzziness is not a *risk source*, unlike the aleatory and the epistemic ones, but it is a *risk factor*.

In the next section, we propose a framework that incorporates, almost, all uncertainties. We believe incorporating those uncertainties in one framework will result in a better understanding about the complexity of problems, especially in S&D context, besides improving the result of analysis and design.

4 Toward a Generic Framework for Secure and Dependable Risks Assessment

A framework is a basic structure that is used to solve or to address complex issues. Ideally, a framework is composed of models, analysis techniques, a process to develop and analyse models, and tools (optional) to assist the framework users. In this paper, we present our revised modelling framework, namely the Goal-Risk framework [27], that has been developed for a general risk analysis. We intend to improve the GR framework to make it more suitable for S&D systems. Moreover, It allows us to capture “most” of uncertainties and not solely analyses the technical systems. However, the wisdom of risk management is how to suppress risks that prevents organizations/enterprise in generating values [20]. Therefore, our proposal aims to provide analysts in managing

S&D risks at the organization level by considering both aspects socio and technical systems. By means of the model, analysts and stakeholders are able to collaborate easily in managing risk.

Example 10. Controller supervisors suppose to be aware about the possible operational risks of the system integrity. However, S&D analysts know about the reliability of radar systems, but they are not well informed how the organization (i.e., Air Traffic Service Provider) perceived possible failures, especially in terms of possible cash-flow disturbance or any legal issues that the organization may face.

One may argue that S&D risks is the “usual” risks in an enterprise (e.g., financial risk, legal risk, environmental risk, operational risk, image risk). In our experiences, S&D risks are, indeed, alike with those risks. One particularity is security risks adhere to malicious intent which rarely considers in the conventional risk analysis - especially in operational risks. Most of S&D risks are exposed to three types of uncertainty as mentioned before, and they trigger other “usual” risks. Moreover, some S&D risks may not emerge as visible phenomena until an exhaustive audit conducted (e.g., Enron company scandals), or even they emerge as a result of the system functionalities (e.g., John Rusnak scandals [28]). Finally, there is no absolute such secure and dependable system because it is hardly possible and makes the system so expensive or hard to use. Therefore, the framework should allow analysts to perform a trade-off analysis over several criteria they are interested.

4.1 Metamodel for the S&D Risk Model

Surveying different risk management approaches across domain applications (e.g., safety-critical [12, 5, 9], security-critical [29, 30], financial [31], enterprise management [20, 32]) we came up with the metamodel depicted in Fig. 2.

Starting from *risk*, we define it as an uncertain event that produces (negative) impacts to assets [7]. In our framework, we extend the notion assets beyond resources that are valuable for the organization. In [33], we categorize risks into three categories: risks threaten resources, risk threaten business processes/tasks, and risk obstructs the achievement of strategic objectives/goals. Events are perceived as normal circumstances that occur in a given time and place [34] or failures in a system. Failures might occur due to malicious intents - attacks (e.g., wiretapping) or just accidental actions - accidents (e.g., filling the wrong values). Basically, an attack (or accident) is a threat (or incident) that exploits (or activates) vulnerabilities/faults in the system. If we can fully control (i.e., protect or remove) all the system’s vulnerabilities, then we may reach the “absolute” security and dependably [13]. The main differences between a threat and an incident are a threat should have three components: attacker(s)/agent(s), motivation(s), and means(s) while an incident does not necessarily have all those components.

Besides attackers, we consider other three relevant roles: risk owner, risk source, and risk controller [35, 20]. For a particular risk, these roles are not necessarily played by same actors³. Essentially, risk owners are affected actors in case a failure occurs.

³Actors [36] includes both human agents and technical systems

S&D Risk Concepts	GR Constructs
Asset	Goal, Task, Resource
Vulnerability	Inherent in Goal, Task, Resource
Risk	
Impact	Impact Relation
Event	Event
Threat	
Attacker	"Bad" Actor
Motivation	"Anti" Goal
Means	"Malicious" Task
Treatment	Task
Mitigation	Alleviation or Contribution Relation
Risk Owner	"requester" Actor
Risk Source	"provider" Actor or "trigger" actor
Risk Controller	"controller" Actor

Table 1: Mapping S&D Risk Metamodel and GR Metamodel using the GR modelling framework in Tab. 1.

Unlike other works based on Tropos/i* [39, 30, 40] that model security as goals, the GR models security and dependability as properties (i.e., reliability, confidentiality, and integrity) of each concept which is needed to be satisfied. In this way, the difference between stakeholders' interests and S&D properties is more apparent and not because the "label" refers to functionality or S&D.

Essentially, a GR model is composed of three conceptual layers: asset, event, and treatment layer. The asset layer captures "things" that are able to generate values for the actor which are depicted in terms of goals, tasks, and resources. *Goals* (depicted as ovals) represent the objectives that actors intend to achieve. *Tasks* (depicted as hexagons) are course of actions used to achieve goals or treat events. *Resources* (depicted as rectangles) are artefacts that are required to achieve goals or to perform tasks. Moreover, resources can be resulted from the execution of tasks. The event layer depicts uncertain events that can affect the asset layer⁴. The treatment layer represents a set of additional measures, depicted as tasks, to mitigate risks (i.e., their likelihood or impacts).

Initially, we only concentrate on assessing the risk of assets' reliability. It aims at ensuring assets will operate correctly to achieve the top goals of each stakeholders. The scene presented in Example 11 are captured by a GR model as depicted in Fig. 3. In that setting (i.e., the risk of unreliable Flight Plan), *Controller* is the one that own this risk. Typically, the decision on whether treating risks or not is driven by the owner. In this setting, *supervisor* acts as risk controller which is responsible to confirm new entries of flight plan by aiming at reducing the likelihood of mis-entry of "null" flight ID. The risk can be triggered by an attack launched by an attacker or simply accidents (e.g., Random Failure FDP Server or Mis-entry of "null" flight ID).

Essentially, both uncertain events activate/exploit the vulnerability at flight plan⁵.

⁴The GR model allows us to model risks (events with negative impacts) and opportunities (events with positive impacts)

⁵In [41], the authors presented the way to represent explicitly the vulnerability by annotating the resource

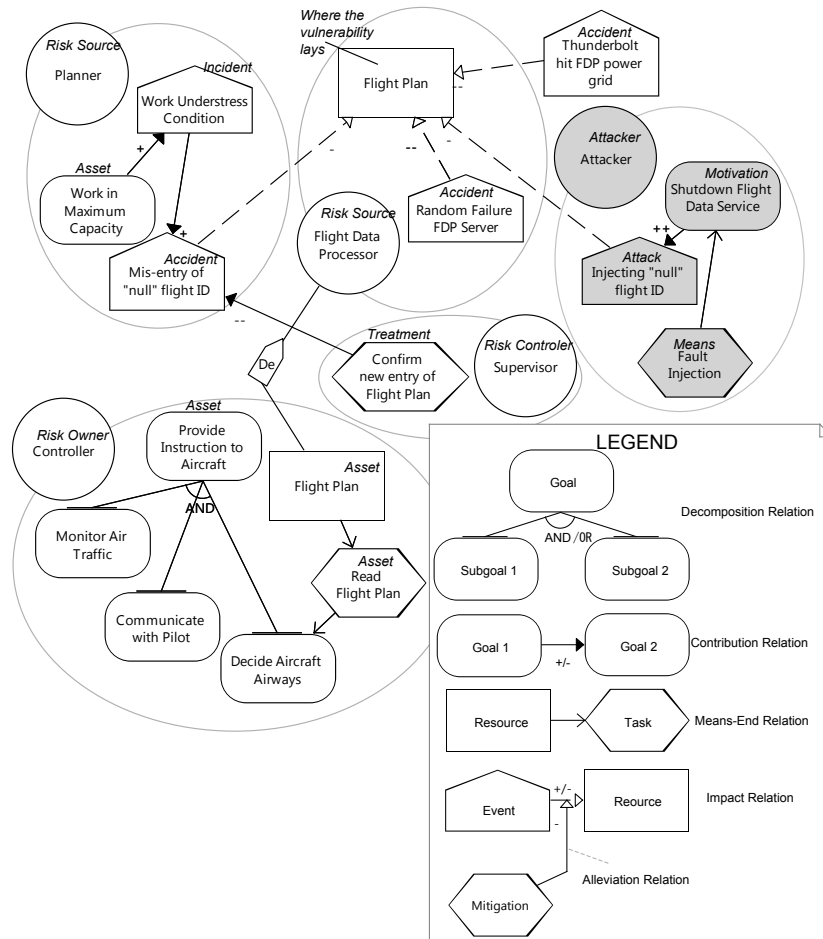


Figure 3: The GR model in Capturing Reliability Risks

Attacks are characterised by attackers, motivation (shutdown Flight Data Processor), and means Fault Injection). In contrast, accidents happen without any motivations; trigger for normal activities (e.g., mis-entry of “null” flight ID) or random events (e.g., random failures of FDP server, Thunderbolt hit the FDP power grid). Notice there **should not be any “positive contribution”** from the actor’s goals/tasks to incidents. Indeed a goal (e.g., working in maximum capacity) can increase the likelihood of an accident, but it is not a direct contribution. There must be an intermediate event (in ATM domain is called by incident) in between. As far as there is a mechanism that suppresses the occurrence of the incident, then the goal achievement will not increase the accident likelihood.

In the context of assessing confidentiality and integrity, we use a similar approach taken for the reliability. Those attributes are represented as a tag (e.g., “I” or “C”) for the asset layer as depicted in Fig. 3. Unlike reliability, not all assets are confidential/integrity-

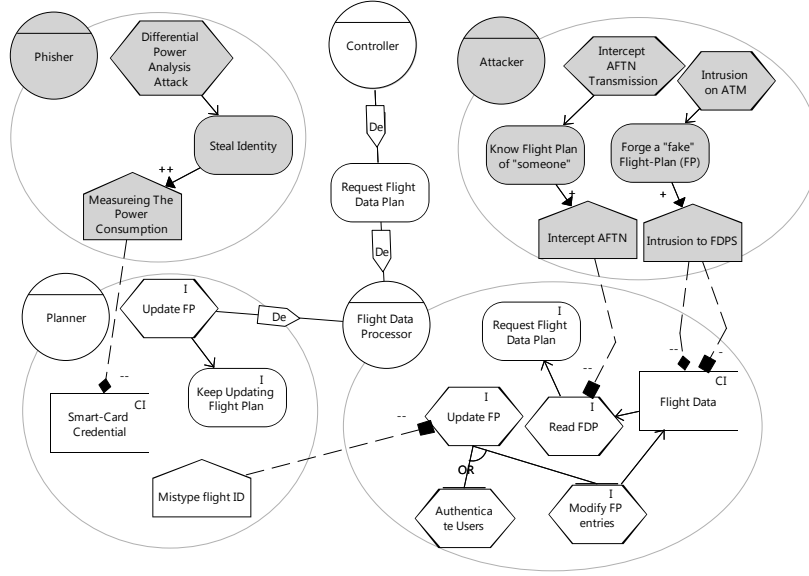


Figure 4: The GR model in Capturing Confidential and Integrity Risk

sensitive. This tagging process can be done by stakeholders and later refined by analysts. Likewise Impact relation, an integrity impact is depicted using “dash-line ending by box”, and a confidential impact is depicted by “dash-line ending by diamond”. Both relations also indicate their severity with a label $(-, --)$.

Confidentiality risks only threaten resources (not goals and tasks) while integrity risks can expose all types of assets. Moreover, confidentiality are prone to risks resulted from attacks, while an integrity breach can be resulted from the normal operation.

Example 12. The event *mistype flight ID* may result in the breach of the flight data integrity. However, the event might be caused by a normal operation of a planner who is careless.

By means of this model, ones can model S&D risks particularly reliable, confidential, and integrity. Each quality is characterised by the likelihood and total expected loss⁶ However, those three qualities are exposed to the three types of uncertainty explained in Section 3. To represent such uncertainties, each constructs are quantified in terms of its evidence (supporting, opposing) to be correctly operated adapting from the *Dempster-Shafer Theory of Evidence*.

$$Sat_i + Den_i + X_i + Y_i = 1; i \in Graph_nodes^7$$

For the complete calculus of this mathematical model, readers may read [42]. In contrast with *Probability Theory*, here the opposing evidence (**denial**) cannot be calculated from the supporting evidence (**satisfaction**) (i.e., $den_i \neq 1 - sat_i$). Moreover,

⁶Notice that each assets has a value. Loss is the sum-up of the loss of value.

⁷ $Sat_i \geq 0; Den_i \geq 0; X_i \geq 0; Y_i \geq 0$

we model the *lack of knowledge*- X_i and *conflicting of knowledge*- Y_i that are required for epistemic uncertainty.

If an event is only exposed to the aleatory nature, we can assume that there is no lack and conflicting knowledge (i.e., $den_i = 1 - sat_i$). For epistemic uncertainty, analysts do not have complete knowledge about the event therefore $sat_i + den_i \leq 1$ and the reminder ($1 - sat_i - den_i = X_i$) can be perceived as the lack of knowledge. By means of the automated reasoner (adapted from [19, 38]), the model obtains the final evidence values. Principally, each relation in the GR model propagates the evidence from source nodes to final nodes following defined semantics [38, 42]. At final, each node has the final evidence for each attributes.

Example 13. Based on Fig. 3, initially the attack injecting “null” flight ID has 50% sat evidence that it will be launched. At the end, the attack injecting “null” flight ID has $sat = 0.4$, $den = 0.5$, and $Y = 0.1$.

Notice at initial analysts assesses the **sat** only and not talking about denial of the attack. Therefore, ones may assume that the analysts do not have any information for the reminder ($X = 0.5$). The example 13 indicates that at the end the **sat** is lower than the initial one. It might be due to the fact that the attacker is not “really” motivated to shutdown flight data service ($den=0.6$). Since it is impossible to have the sum of evidence bigger than 1, one may assume there is conflicting evidence ($Y = 0.1$).

Based on those final values, ones may compute how much the likelihood (λ) of “things” occurs (i.e., goals to be achieved, tasks to be executed, resources to be provided, and events occur). In this framework, if an event is exposed only by aleatory then $\lambda = sat$. If events (and also goals, tasks, resources) are exposed in epistemic one (i.e., $X \neq 0$ or $Y \neq 0$) then the likelihood is a imprecise value. In this framework, we define it as a range:

$$bel_i \leq \lambda_i \leq pla_i^8; i \in Graph_nodes$$

$$bel_i = \begin{cases} Sat_i, & \text{if } i \text{ is a risk/bad thing;} \\ Sat_i - \llbracket X_i - Y_i \rrbracket, & \text{otherwise}^9. \end{cases}$$

$$pla_i = \begin{cases} Sat_i + X_i + Y_i, & \text{if } i \text{ is a risk/bad thing;} \\ Sat_i, & \text{otherwise.} \end{cases}$$

Essentially, the belief bel_i assumes at the “at least” value for evidence while the plausibility pla_i assumes at the “most” value one. By means this representation, ones can see the (un)fuzziness of the evidence of a goal and decide how much mitigation are required accordingly.

⁸It is specified following the similar notion on **belief** and **plausibility** in the Dempster-Shafer Theory [43]

⁹ $\llbracket x \rrbracket = X$ if $x \geq 0$, otherwise $\llbracket x \rrbracket = 0$

By means of the model and its formalization, analysts can understand what are risks associated to the system and assess them. Moreover, they can also understand how to reduce the uncertainty, according to its nature, of risk and how to mitigate the risk [27]. For instance, if a risk (or an opportunity) has a wide-range (imprecise) of likelihood (e.g., 20%-70%) then it is better if the treatment is elicited based on the biggest (smallest) value. In the case of incomplete knowledge of an event, analysts can perform some empirical efforts (e.g., interviews, expert judgements) to reduce the lack of knowledge or to minimise the conflict of knowledge about the event. Consequently, these efforts may reduce the uncertainty about the event. Imprecise uncertainty is the only type that may not result in risk. It depends on the sensitivity of the system towards the imprecision aspect.

5 Concluding Remarks

In this paper, we have discussed about basic concepts of security and dependability proposed in literature and how they are related to uncertainty. We have explained how S&D quality attributes can be reduced to reliability, confidentiality, integrity, and recoverability qualities. We have proposed a modelling framework that captures critical concepts of having a S&D system (e.g., assets, attack, accident, failures, treatment) and reason about related risk from three different dimensions (namely, reliability, confidentiality, and integrity). The nature of uncertainty is used along the assessment process and is used to find the most appropriate treatment for risk mitigation. Moreover, one may perform a cost-benefit analysis [44] over a GR model to maximise the effectiveness of cost for countermeasures in achieving those three attributes using the forward reasoning [38]. For enrichment, trade-off analysis [41] can be conducted using a GR model to see the trade-off among attributes for a given set of treatments.

Currently, our approach is able to capture and analyse some parts of availability property (i.e., reliability, recoverability) [33]. However, it cannot deal with the interruption (e.g., Denial-of-Service) because in this case the system runs correctly (reliable) but it is not accessible by end-users. Our model elicit “necessary” evidence for the availability of a system, but further analyses (i.e., responsiveness) are required to ensure the availability. In this paper, we model and analyse necessary conditions to be secure and dependable, but unfortunately it is not, surely, sufficient (i.e., if there both conditions are different). In fact, it is hardly possible to have sufficient evidence/condition to judge whether the system is secure or *not*. Security engineering, often, conducts with lack of knowledge. For instance, our firewall can filter all malicious traffic now, but not necessarily the case for next week. It is due to the nature of the quality attributes which their breaches are not necessarily appeared by users and always increase over the time.

The value of X or Y refers to the incomplete knowledge of the analysts about the subject. Ideally, they should try to reduce those values using some techniques as indicated in the Section 3. However, those techniques cannot be counted as mitigation. It is aiming at increasing the preciseness of the assessment, and at the end it may reduce the possibility of “over-shoot” in adopting countermeasures (i.e., reduce the cost of countermeasures).

Analysts should try to remove the vulnerability that lay in the system. In [45, 37, 3],

the authors proposed structure ways to identify vulnerabilities and possible removals. However, we realise it is impossible to remove all vulnerabilities. Therefore, a set of means is necessary to mitigate risks: prevention, detection-recovery, alleviation, and restoration. Prevention means aims at preventing the exploitation/activation of those vulnerabilities. Detection-recovery means aim at detecting the present of intrusions or errors and try to recover the system from them. Alleviation intends to reduce the severity of failures resulted from attack or accident. Finally, analysts should provide means to perform restoration after failures. Here, we distinguish between restoration and recovery. Restoration recovers the system from failures while recovery is aiming at recovering from errors. However, in many cases they are alike. In some setting, we may transfer the risks into the third party (e.g., insurance companies) which can be seen as part of the mitigation strategies.

As mention in the previous section, risk controller and risk owner can be different actors. It might be the case that the controller does not perceive a risk as greater as the owner. This situation may result in agreed countermeasure are not executed because the controller perceives less risky than the one that owner does. Therefore, further analyses are required to ensure that the perceived risks do not significantly deviate from the actual ones.

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