

Requirements Quality Assurance in Industry: Why, What and How?

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Abstract. *Context and Motivation:* Natural language is the most common form to specify requirements in industry. The quality of the specification depends on the capability of the writer to formulate requirements aimed at different stakeholders: they are an expression of the customer's needs that are used by analysts, designers and testers. Given this central role of requirements as a mean to communicate intention, assuring their quality is essential to reduce misunderstandings that lead to potential waste. *Problem:* Quality assurance of requirement specifications is largely a manual effort that requires expertise and domain knowledge. However, this demanding cognitive process is also congested by trivial quality issues that should not occur in the first place. *Principal ideas:* We propose a taxonomy of requirements quality assurance complexity that characterizes cognitive load of verifying a quality aspect from the human perspective, and automation complexity and accuracy from the machine perspective. *Contribution:* Once this taxonomy is realized and validated, it can serve as the basis for a decision framework of automated requirements quality assurance support.

Keywords: Requirements engineering · Requirements quality · Natural language processing · Decision support

1 Introduction

The requirements engineering process and the artefacts used in coordination and communication activities influence the performance of downstream development activities [6]. While research has proposed myriads of formal, semi-formal and informal methods to convey requirements, plain natural language (NL) is the *lingua franca* for specifying requirements in industry [14, 17]. One potential reason is that NL specifications are easy to comprehend without particular training [3]. However, NL is also inherently imprecise and ambiguous, posing challenges in objectively validating that requirements expressed in NL represent the customers' needs [1]. Therefore it is common practice to perform some sort of review or inspection [14] to quality assure NL requirements specifications. While there exists a plethora of methods to improve requirements specifications [15], there are no guidelines that would support practitioners in deciding

which method(s) to adopt for their particular need. We think that a first step to such a decision framework is to characterize the means by which quality attributes in requirements specifications can be affected. Therefore, we initiated an applied research collaboration with the Swedish Transport Administration (STA), the government agency responsible for the rail, road, shipping and aviation infrastructure in Sweden. STA’s overall goal is to improve the communication and coordination with their suppliers, mostly handled through NL requirements specifications. Infrastructure projects vary in duration (months to decades) and budget (up to 4 Billion USD), requiring an adaptive quality assurance strategy that is backed by methods adapted to the needs of the particular project. The large number of requirements (several thousands) and the need to communicate them to various suppliers makes specifications in NL the only viable choice. Still, STA needs to quality assure the requirements and decide what level of quality is acceptable. In this paper we present the basic components for a taxonomy that will drive, once the research is completed, a requirements quality assurance decision support framework. To this end, we illustrate a research outline aimed at answering our overall research question: **How can we support practitioners in achieving “good-enough” requirements specification quality?**

2 Related Work

Davis et al. [7] proposed a comprehensive set of 24 attributes that contribute to software requirements specification (SRS) quality. Saavedra et al. [16] compared this set with later contributions that studied means to evaluate these attributes. Similarly, Pekar et al. [15] reviewed the literature and identified 36 studies proposing techniques to improve SRS quality. While Agile software development is notorious for promoting as little documentation as possible [10], Heck and Zaidman [13] identified 28 quality criteria used for Agile requirements, six of them being novel and specifically defined for Agile requirements. All these reviews point to relevant related work potentially contributing to the components of a decision support framework for requirements quality assurance. The importance of providing decision support to practitioners is growing hand-in-hand with the complexity of today’s developed software products and the available number of technologies to realize them [12]. To the best of our knowledge, no framework exists to support the selection of requirements quality assurance techniques.

3 Characterizing Requirements Quality Assurance

The purpose of this taxonomy is to characterize the components that are involved in the process to achieve a particular requirements quality (RQ) level (Fig. 1). This systematization then allows to take informed decisions about effort and potential impact for RQ improvement.

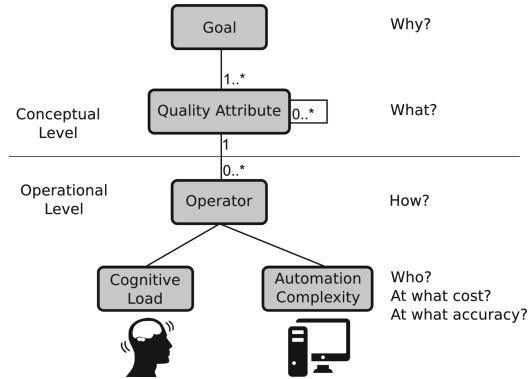


Fig. 1. Requirements quality assurance taxonomy

A *goal* determines what the improvement of RQ should achieve. Typical goals could be to improve the communication between stakeholders, to improve the ability to verify the product, or better cost estimates. Different goals can also contradict each other. Goals are important as they provide a scope that limits the potential actions on the operational level to a set that is economically acceptable - this enables focus of efforts to assure certain quality aspects within the given opportunities of the resources afforded.

Quality attributes describe the favourable properties of a requirement. For example, unambiguity is commonly defined as the quality of a statement being interpretable in a unique way. Quality attributes for requirements have been described in numerous quality models, reviewed by Saavedra et al. [16]. Quality attributes are not independent, i.e. one attribute can positively or negatively influence another. Figure 2 provides an overview of RQ attributes and their relationships to each other. For example, atomicity positively influences design independence, traceability and precision of a requirement, as indicated by the (+) in Fig. 2. On the other hand, unambiguous requirements, often achieved by higher formality, are generally also less understandable.

Goals and quality attributes build the *conceptual level* of the taxonomy. They can help to answer questions pertaining to why an improvement of RQ is necessary, and what quality attributes are associated with that goal. Taking the example from earlier, improving the ability to verify the product based on the stated requirements, one can see in Fig. 2 that many quality attributes influence requirements verifiability. Depending on constraints in the operational level, discussed next, one can decide how to reach the stated goal by choosing a set of quality attributes, which in turn are associated with operators.

Operator is the generic term we use for instruments that tangibly characterize quality attributes. An operator provides a definition of how a requirement is analysed w.r.t. the associated quality attribute. Examples of operators are metrics [8, 11], requirement smells [9] or rules and constraints on how to formulate requirements. An operator can be implemented by either a person or a computer program (or both). In either case, we want to characterize the operator

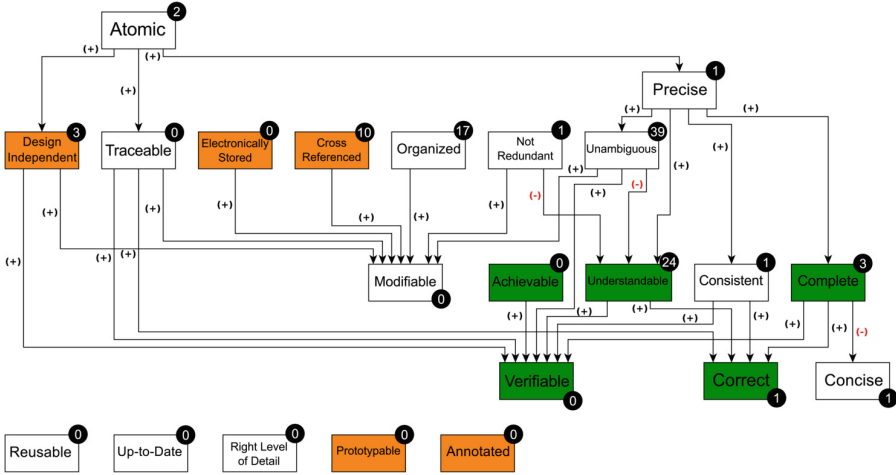


Fig. 2. Quality attributes and their relationships (adapted from Saavedra et al. [16]; color coding and numbers are our addition, and used and explained in Sect. 4)

by some notion of cost and accuracy, providing input for the decision on how and whether at all to realize the operator. We borrow the concept of *cognitive load* from the field of instruction design where cognitive load theory [18] is used to describe and improve learning efficiency. Each operator is associated with a level of intrinsic cognitive load, describing the complexity of applying the operator on a single requirement or a complete specification. For example, if the operator is the ambiguous adverbs requirements smell [9], then the intrinsic cognitive load is determined by the number of ambiguous terms one has to remember to detect these terms in the requirements text. Since cognitive load is additive [18], there are (individual) limits to the efficiency of applying operators, and is therefore one determinant for the effective cost of RQ assurance. If an operator is realized through machine-based processing of information, we characterize this realization by its *automation complexity*. Continuing with the example of ambiguous adverbs, the automation complexity of this operator is low as it can be implemented with a dictionary [9]. On the other hand, some of the requirements writing rules found in STA are rather complex. For example, one rule states that repetition of requirements shall be avoided and a reference to a general requirement shall be made (addressing redundancy). The detection of rule violations requires the analysis of the complete specification, identifying similar phrased statements. While this is certainly possible (e.g. with code clone and plagiarism detection [5]), the analytical complexity is higher than for a dictionary lookup.

4 Research Outline

The taxonomy serves three main purposes which are outlined in this section, together with six research questions and our planned approaches to answer them.

4.1 Prioritize Quality Attributes

We have asked six requirements experts at STA to rank RQ attributes (definitions were extracted from the review by Saavedra et al. [16]) by their importance using cumulative voting [2]. Figure 2 shows the five top and bottom attributes in green and orange respectively. Individual quality attributes have been researched earlier, focusing on ambiguity, completeness, consistency and correctness [15]. While the perceived importance of completeness and correctness is matched by research on these attributes, ambiguity and consistency were ranked by the experts only at position 13 and 16 respectively. At first sight, this might indicate that research focus needs adjustment. However, taking into consideration the relationships between quality attributes, we see a moderate overlap between the needs at STA and past research. Nevertheless, there are certain quality attributes whose evaluation has seen little research, like traceability [15], while being important for STA since they affect verifiability and correctness. The relationships between quality attributes inform us also about potential inconsistencies among the goals of quality improvement. For example, design independence was ranked by STA's experts on position 21 while it affects verifiability, ranked at position 3. This could indicate that, while verifiability is important for STA, design independence as a related aspect has been overlooked as a means to achieve this. These examples show how the relationships between quality attributes can be used to analyse the goals of the company. However, since Saavedra et al. [16] deduced the relationships shown in Fig. 2 by interpreting the quality models they reviewed, these dependencies need further empirical validation, leading to *RQ1: To what extent do requirements quality attributes affect each other?* One approach to address this question, dependent on the answers to the questions in Sect. 4.2, would be to analyse the correlation between operators for different quality attributes. We plan to perform this analysis at STA, which in turn partially answers *RQ2: To what extent can quality attribute rankings be used for planning quality assurance activities?* Further inquiries at STA are needed to identify factors that affect planning, such as timing (does quality attribute importance depend on the project phase?) and implementation cost.

4.2 Determine Operators and Their Accuracy

At STA we have identified 110 *operators* in the form of requirements writing rules. These rules describe how requirements shall be formulated and provide review guidelines. Table 1 shows five examples of writing rules. We have mapped, where the description allowed it, which quality attribute was primarily targeted by each rule. The numbers in Fig. 2 indicate how many operators we identified for each quality attribute. Several quality attributes have no or very few associated operators, leading to the question *RQ3: Which quality attributes can be characterized by an operator?* We plan to answer this question by systematically reviewing the literature, extending the work by Saavedra et al. [16], Pekar et al. [15], and Heck and Zaidman [13]. On the other hand, we have identified

110 operators in STA, leading to the questions *RQ4: How can NL processing be used to implement operators?* and *RQ5: What is the accuracy of these operators in relation to state-of-practice?* We estimated that 40–50% of the writing rules in STA can be implemented with current techniques, e.g. as proposed by Femmer et al. [9]. However, as indicated in the last column of Table 1, techniques to implement rules 4 and 5 still need to be determined. In addition we plan to evaluate the practical benefits of machine-supported RQ assurance compared to the state-of-practice, i.e. manual quality assurance, at STA.

Table 1. Examples of requirements writing rules at STA

Rule	Quality attribute	Implementation
1. No time should be specified in the technical documents. Instead, refer to the Schedule document	Non-redundant	Named entity extraction
2. Numbering of figures, illustrations and tables should be consecutively numbered throughout the document, starting from 1.	Organized	Document meta-data analysis
3. Numbers “1–12” shall be written as shown in the following example, “to be at least two (2).”	Unambiguous	POS Tagging
4. Terms such as “user”, “dispatcher”, “operator” should be used consistently	Unambiguous	TBD
5. If a functional requirement is supplemented by additional requirements to clarify fulfilment, these must be written as separate requirements	Atomic	TBD

4.3 Estimate Cognitive Load and Automation Complexity

Applying all 110 operators on a specification consisting of thousands of requirements is a cognitively demanding task. For deciding how to implement an operator, it would be useful to be able to estimate the cognitive load each operator will cause and the complexity to implement the operator in a computer-based support system, leading to *RQ6: How can the cognitive load and automation complexity of an operator be estimated?* Cognitive load could be approximated by a heuristic that describes whether the application of the operator requires domain knowledge or not, and to what extent context needs to be considered. Context could be defined as “local”, referring to a single requirement, “regional” referring to a section or chapter in the specification, or “global” the whole specification and beyond, e.g. regulations and standards. There exist also multiple approaches to measure cognitive load directly [4]. Automation complexity could be estimated by categorizing operators on the linguistic aspect they address. Operators that require semantic understanding are more complex than operators that require syntactic or lexical analyses of a requirement. The least complex operators are statistical, i.e. analyses that work with letter, word or sentence counts. Since,

to the best of our knowledge, no such characterization of operators exists, we plan to collaborate with experts from both neuropsychology and linguistics to perform literature reviews and design experiments.

5 Conclusion

In this paper, we have proposed a requirements quality assurance taxonomy that, once the stated research questions are answered, forms the engine for a decision framework that allows companies to initiate or improve their requirements quality assurance program through (a) realizing the consequences of dependencies between quality attributes in their current manual activities for quality assurance, (b) mapping cognitive load to the prioritized actions for quality assurance, and (c) enabling the decision on the trade-off between manual and machine-supported quality assurance, given cost and accuracy of the choices.

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