

Visualization of the Values of Requirements Attributes with Goal Models

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Abstract. *Attributes such as cost, risk, priority, and customer satisfaction are integral components of requirements specifications, helping with decision-making during the software development process. Moreover, requirements status and stability are relevant meta-information for requirements management. Even though goal modelling is a visual approach for requirements engineering that enables expressing and reasoning about goals, qualities, tasks and resources, existing approaches do not include the aforementioned attributes as part of the visualization of the models. This paper presents a set of strategies for visually representing requirements attributes over goal models, aiming at minimizing the cognitive effort required in their analysis. The proposal is supported by a tool that implements the visualization strategies here defined. Empirical evaluation of the effectiveness of the proposal, however, is expected to be carried out as future work.*

1. Introduction

Requirements attributes are integral components of a requirements specification, complementing the information provided by requirements statements and user stories. For instance, Robertson and Robertson [Robertson and Robertson, 2006] argue that natural-language statements are insufficient. So, they propose the use of additional attributes: Requirement number, Type, Related event/use case, Description, Rationale, Originator, Fit Criterion, Customer Satisfaction/ Dissatisfaction, Priority, Conflicts, Supporting Material and History. Other proposed requirements attributes include Acceptance Criteria, Author, Cost, Date, Dependency, Difficulty, Identification, Origin/source, Owner, Priority, Release Number, Rationale, Responsible, Risk, Source, Stability, Stakeholder Priority, Status, Subsystem, Type, Value, Verification Method, and Version Number [ISO/IEC/IEEE, 2011] [Wiegers and Beatty, 2013].

In the context of goal modelling, some tools already enable the specification of such attributes for the elements of a model. For instance, jUCMNav [Amyot et al., 2011] is an Eclipse-based tool for the User Requirements Notation, which comprises not only a goal-based language (Goal-oriented Requirement Language – GRL) but also the Use Case Maps notation. We can detail the elements of a GRL model through the following pre-defined attributes: Name, Description, Criticality, Importance, and Priority. Moreover, custom attributes can be defined by users through a Metadata panel. Figure 1-a shows an example of a task with four added attributes: Benefit, Cost, Penalty, and Risk. The piStar tool [Pimentel and Castro 2018] [Pimentel et al. 2019] is a web-based tool for the iStar 2.0 notation [Dalpiaz, Franch and Horkoff 2016]. In this

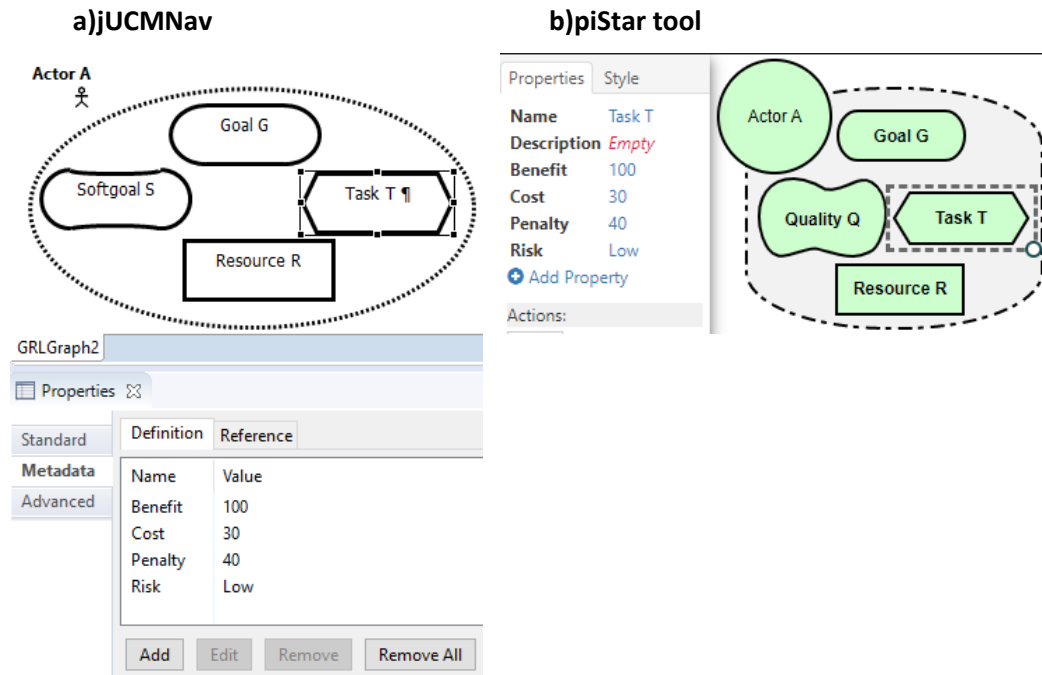


Figure 1. User-defined attributes in two goal modelling tools: jUCMNav (a) and piStar tool (b).

tool, the i^* elements have two pre-defined attributes: Name and Description. Like in jUCMNav, users can define custom attributes through a Properties panel, as exemplified in Fig. 1-b.

Since the visualization of such attributes is local to the element – i.e., only the attributes of the selected element are displayed – it is difficult to perform a global analysis of the model considering the indicated attributes. Some analyses that could be made, with a broader visualization of the requirements attributes within a goal model, include: “Which requirements are the riskiest?”, “Do related requirements have similar stability values?”, “Are the highest priorities requirements concentrated on some actors?”, “Which requirements have not been implemented yet (status)?”, “Are the priorities of tasks consistent with the priorities of the quality goals they contribute to?”.

Visualization comprises the transformation of raw data for easier assimilation through sight, increasing consciousness and insight on the data [Gotel, Marchese and Morris 2007]. It is well known that visual artefacts facilitate thought [Card, Mackinlay and Shneiderman 1999]. Particularly, the beneficial role that visualization strategies can play in the context of Requirements Engineering has been extensively documented [Cooper et al. 2009] [Abad, Noeen and Ruhe 2016].

In this paper, we present a tool-supported approach for globally visualizing the numerical and categorical attributes associated with elements on a goal model – namely, i^* models. The choice of the i^* framework [Yu et al. 2011] as the baseline is due to its expressiveness in modelling and reasoning about early requirements and social networks. In particular, i^* models express not only system requirements but also stakeholders’ goals and dependencies, as described in the next section.

Examples are given in the domain of requirements prioritization and release planning. Prioritization is a key activity of the requirements engineering process, enabling stakeholders to identify what are the most important features of the system-to-be [Berander and Andrews 2005] [Achimugu et al. 2014]. Several studies have already provided significant contributions on this topic. For instance, many relevant prioritization criteria that can be adopted in order to inform the prioritization process have already been identified and classified – e.g., customer satisfaction, business value, risk and implementation effort [Riegel and Doerr 2015]. Moreover, several requirements prioritization techniques can be used to guide such prioritization [Bukhsh, Bukhsh and Daneva 2019]. However, as evidenced by [Cavalcanti et al. 2018], there is a lack of visualization approaches supporting requirements prioritization.

Section 2 introduces the main concepts of *i**, which is the modelling framework upon which this proposal is built. Section 3 presents the visualization strategies that are proposed in this paper. Tool support for the proposal is described in Section 4. Related work is discussed in Section 5. Lastly, the paper is concluded in Section 6.

2. Baseline – *i** Framework

The *i** (i-Star) Framework [Yu et al. 2011] was chosen as the baseline for this proposal. In particular, the iStar 2.0 standard [Dalpiaz, Franch and Horkoff 2016], which aims at defining a consistent and clear set of core concepts of the *i** language, was adopted as the basis for this proposal.

To provide an overview of the *i** language, we adopted the example of an online store system (Figure 2). The main goal of this system is to have the “Internet Shop Managed”. In order to do so, it needs to handle item searching (goal) and internet orders (goal), as well as allowing to update the catalogue of products (task). Item searching can be handled by querying the database (task) or by consulting the catalogue of products (task). Internet orders can be handled using a Fill Secure Form Order (task) or a Fill Standard Form Order (task). Moreover, the system is expected to be attractive to new consumers, as well as providing security and availability.

The use of visual models, in contrast with natural language statements, can facilitate the occurrence of insights and the identification of inconsistencies through the explicit representation of relationships between the different elements of the model, as exemplified next.

- **Actors** – by including different actors and their dependency relationships with elements of the System-To-Be actor, it is possible to infer some priorities if one considers the weights of these actors. For instance, in Figure 2 the Catalogue Consulting and Database Querying tasks are directly linked to customer needs. Nevertheless, this reasoning is partial, since not every stakeholder (such as developers and customers) take part in the system’s strategic dependencies.
- **AND-Refinements** – since it is possible to visualize why the system includes each task, one can partially infer the priority of sub-elements. For instance, in Figure 2, “Item Searching Handled” is an and-refinement of “Internet Shop Managed”. Thus, if the latter is (for instance) high priority, the former will likely also be high priority.
- **OR-Refinements** – the information on alternative refinements can aid in release planning. For instance, if an alternative (e.g., Database Querying) has already been

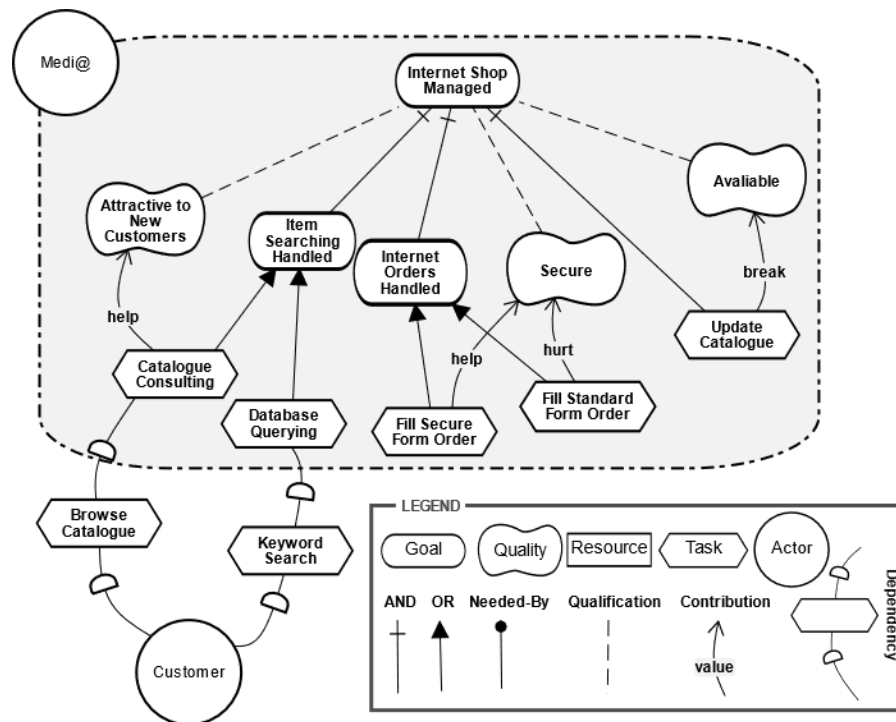


Figure 2. Excerpt of the Medi@ Example

implemented, the other alternative (Catalogue Consulting) may be postponed to a later release, or perhaps even discarded entirely, since the first alternative is sufficient to satisfy the Item Searching Handled goal.

- Contribution links – the contribution towards qualities can aid the selection of alternative tasks. For instance, in Figure 2 there are two options to have Internet Orders Handled: Fill Secure Form Order and Fill Standard Form Order. If we consider the Secure quality, the best alternative would be the Fill Secure Form Order, which helps to satisfice said quality.

3. Visualization strategies

This section presents the visualization strategies originally proposed in [Pimentel and Lencastre, 2020], with the purpose of enhancing requirements visualization for prioritization (Section 3.1) and release planning (Section 3.2).

3.1. Visualization of ranged values

Requirements engineering visualizations can be classified in five categories [Cooper Jr. et al. 2009]: Tabular, Relational, Sequential, Hierarchical and Quantitative/Metaphorical. Regular i* models are able to express relational and hierarchical information, i.e., sets of connected nodes and decomposition of elements, respectively. In [Pimentel and Lencastre, 2020] we have proposed to enrich i* models in order to express prioritization attributes such as importance, cost and risk through quantitative/metaphorical visualizations, i.e., “visual clues such as color, shape, line thickness, and size to convey meaning at a glance” [Cooper Jr. et al. 2009]. Moreover, [Moody 2009] described eight visual variables that can encode information on visual models: horizontal position, vertical position, size, brightness, colour, texture, shape and

orientation. Whereas regular i^* models make use of horizontal and vertical position, as well as shape, our proposal makes use of colour, brightness, and shape.

Specifically, a range from yellow to red is adopted to represent variation on the values of an attribute, from lowest to highest. The specific colours were carefully selected in a way to correspond to a brightness variation, where lower values have higher brightness and higher values have lower brightness. The benefits of this selection are twofold: as redundant coding [Moody 2009], brightness acts as an additional signal that reinforces the meaning of the colour; moreover, it allows to differentiate (to a lesser degree) the attribute values even when the image is seen by people with visual colour deficiencies, as discussed in Section 3.3, even though further empirical evaluation is warranted to confirm this hypothesis.

Even though colour is one of the most effective visual variables, size is the one with higher capacity (i.e., number of perceptible steps), apart from shape [Moody 2009]. Thus, we propose to also use a range of sizes to represent the value of requirements attributes, ranging from small (lower values) to large (higher values).

Besides these three visual variables, a textual encoding was additionally adopted to express the values of a given attribute with precision. This encoding is represented by a numerical value contained within an icon that is attached to an i^* element.

In the remainder of this paper the terms colouring is used to refer to variations of colour and brightness together, resize to refer to variations of size, and icons to refer to the textual encoding. Combination of these variations are possible, both for a single attribute and for multiple attributes. For instance, colouring and resize can be used to represent the range of cost, thus reinforcing the signals regarding a single attribute. Alternatively, one could use colouring to represent cost while resize is used to represent benefit, thus representing multiple attributes (cost and benefit) in the same visualization. Regardless of the strategies chosen, consistency is key to prevent confusion within a team, project, or organization.

Figure 3 shows different visualizations of attributes in an i^* model. Figure 3-a shows a regular i^* model. Figure 3-b shows the same model but now making use of the colouring strategy, where low to high values of importance are represented with colours from yellow to red, respectively. It can be seen, for instance, that Quality A has much higher importance than Task E. In black-and-white (shades of grey) versions of this image, the different values can still be observed, even though it is harder to make such distinction – brighter elements have lower importance, whereas darker elements have higher importance.

Figure 3-c exemplifies the resizing option, where smaller elements have lower importance and larger elements have higher importance. A potential issue here is: what to do with elements that have undefined values for a given attribute? We have chosen to make these elements smaller, so that they are not confused with high-valued elements. However, this may lead to a confusion between low-valued elements and elements with undefined values.

Lastly, Figure 3-d shows the exact importance value of each element by means of a numerical value contained within an icon attached to the element. In this example, these icons are also resized according to importance values – larger circles represent higher importance values.

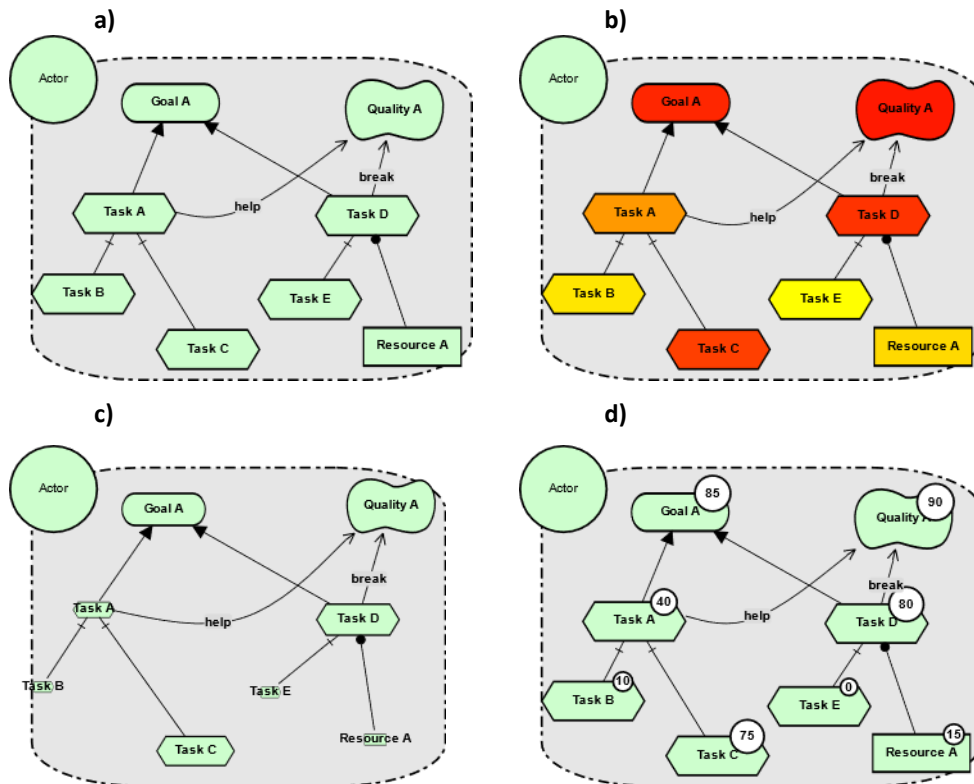


Figure 3. Different visualizations of the same *i** model: a) original; b) colouring elements by importance; c) resizing elements by importance; d) importance value on icons attached to each element, with icons resized according to their value.

In summary, Figure 3-b, Figure 3-c and Figure 3-d represent values of the same attribute (importance) through different visualization options. In this example, it can be observed that: both Task D and Criterion Goal A have high importance, even though Task D breaks Criterion Goal A; Task E has low importance, even though Task E is a part of Task D, which has high importance. Based on these potential inconsistencies it may be decided that the current set of importance values are not satisfactory, hence further analysis is required.

Multiple Attributes. While the attached icons shown on Figure 3-d only present one attribute (namely, importance), many attributes can be displayed at once as well. Figure 4 shows different options for visualizing the cost, benefit, and risk of each requirement (in this order). Proper tool support can enhance this visualization by providing interactive tooltips that display the name of the attributes being visualized.

On Figure 4-a, each icon shows the value of each attribute (cost, benefit, and risk, respectively). Figure 4-b and Figure 4-c show not only the value but also a visual reinforcement: size and colour (respectively). This reinforcement can make it easier to compare each attribute. Figure 4-d shows a combination of the previous options, where the attribute values are also represented by both the colour and the size of each icon.

It is also possible to hide the value of each attribute, relying solely on visual cues to compare the attributes (Figure 4-e, Figure 4-f, and Figure 4-g). In this case, the first letter of the name of each attribute is displayed (C - Cost; B - Benefit; and R - Risk).

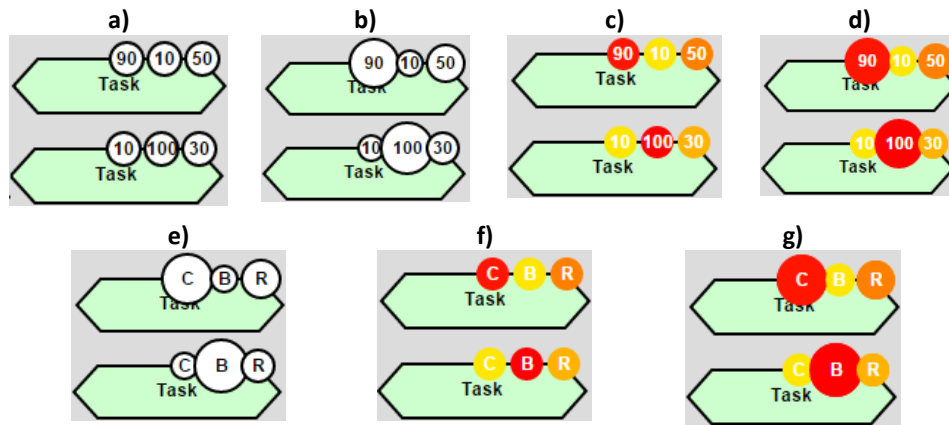


Figure 4. Different options for icons showing, respectively, the Cost, Benefit, and Risk of each attribute. a) plain; b) resizing; c) colouring; d) colouring and resize; e) resizing, attribute first letter; e) colouring, attribute first letter; e) colouring and resizing, attribute first letter.

This option reduces the precision of the visualization but reduces the effort of remembering the names of the attributes that are being visualized.

3.2. Visualization of binary classifications

Sometimes it is useful to have binary classifications related to an attribute, such as: which tasks are assigned a given release; which qualities have importance higher than 80; which requirements have the status “approved”. which Release planning is an activity strongly related to requirements prioritization. We propose three strategies for such visualization: line-decoration, resizing, and colouring. Here, resizing and colouring differ from the strategies described in the previous sub-section by considering only two values, instead of a range of values.

Figure 5 shows different visualizations of the same i^* model, where Task D, Task E, and Resource A are scheduled for the first release. In Figure 5-a, these elements are highlighted by means of thick, dashed lines. In Figure 5-b, the elements that are not included in the first release were made smaller, thus allowing the viewers of the model to focus on the elements that are scheduled for the release under analysis. Lastly, Figure 5-c highlights the elements scheduled for the first release by means of colouring. These different visualization options can be combined, making the differentiation between releases even more distinct.

3.3 Colour vision deficiencies

Colour vision deficiencies are a factor that may compromise the effective use of colour-based visualization. To analyse this issue, we have simulated how the chosen gradients would be perceived by people with extreme cases of colour blindness: protanopia (no perception of red), deuteranopia (no perception of green), tritanopia (no perception of blue), and achromatopsia (no perception of any colour other than shades of grey). Figure 6 presents the simulated images for a green-yellow-red gradient, whereas Figure 7 shows the images for the yellow-red gradient adopted in this paper. This simulation was created using the Developer Tools provided by Google Chrome. Based on this simulation, we have formulated the following hypotheses that should be empirically evaluated:

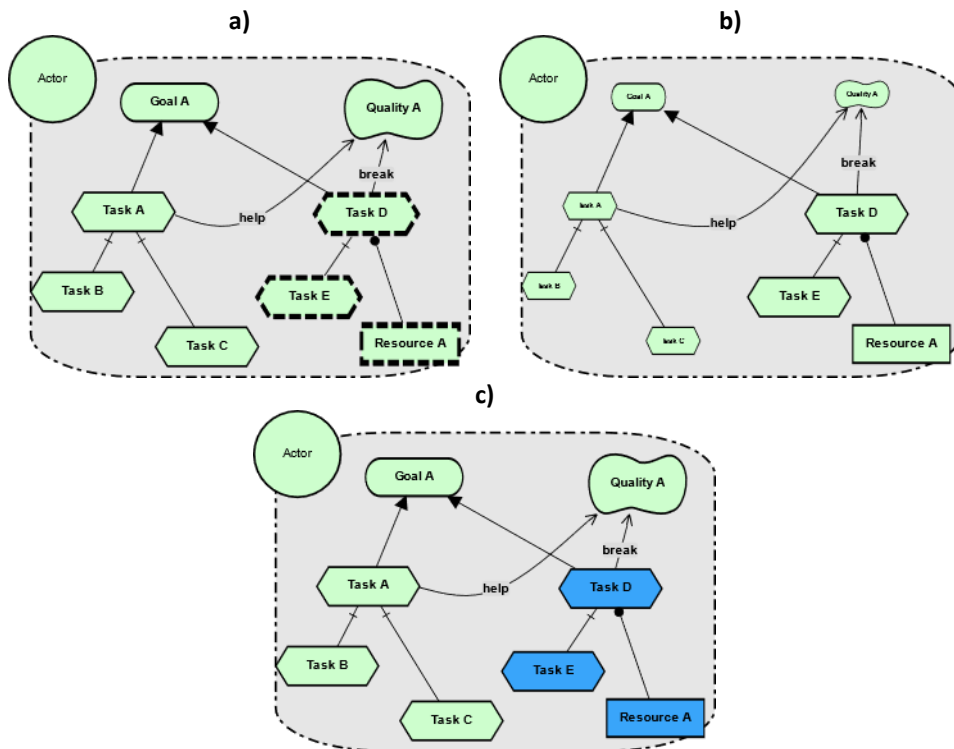


Figure 5. Different options for visualizing release information: a) line decoration; b) resizing; c) colouring.

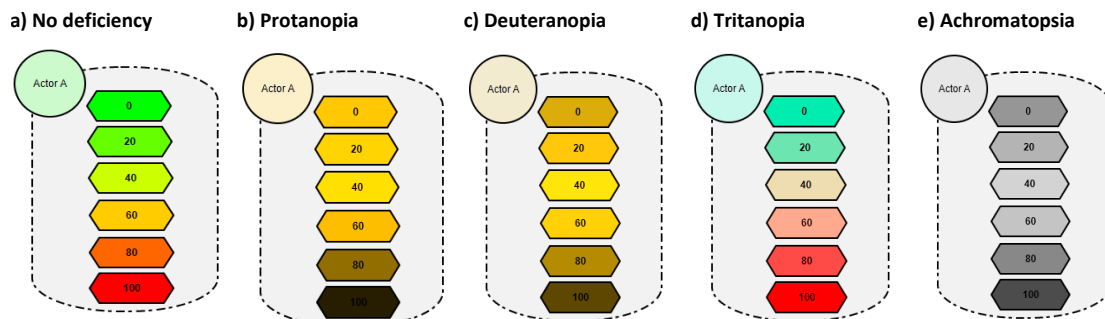


Figure 6. Simulation of colour vision deficiencies on the green-yellow-red gradient

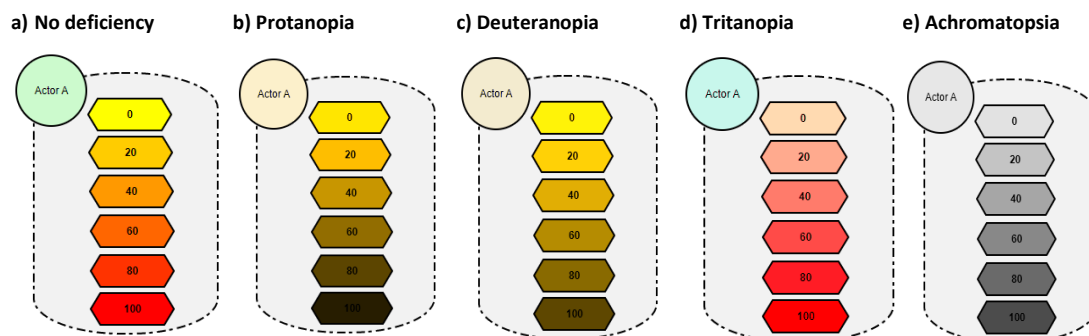


Figure 7. Simulation of colour vision deficiencies on the yellow-red gradient

Hypothesis 1: the green-yellow-red gradient (Figure 6) is more suitable than the yellow-red (Figure 7) gradient for people that are not afflicted with colour vision deficiencies. We believe this is the case since the first gradient provides a more distinctive colour for each step.

Hypothesis 2: the yellow-red gradient (Figure 7) is more suitable than the green-yellow-red (Figure 6) gradient for people afflicted with colour vision deficiencies. This is the case since, for some deficiencies, the perceived colour can be too hard to distinguish at different scale points. For instance, people with deuteranopia (Figure 6-c) and achromatopsia (Figure 6-e) will have difficulty in distinguishing the colours corresponding to the values 0 and 80.

4. Tool support

In order to enable the use of the visualization strategies described in the previous section, we have developed an extension of the piStar tool, a web-based general-purpose i* modelling tool [Pimentel and Castro 2018] [Pimentel et al. 2019]. This extension, available online at <https://www.cin.ufpe.br/~ler/supplement/cadernosime2020/visualization/>, comprises two new functionalities: ranged value visualization, and binary visualization.

The user of the tool (e.g., a requirements engineer), after creating an i* model depicting the system requirements, may need to add some metadata to its requirements, such as: rationale, origin, cost, and benefit. Instead of documenting this metadata in a separate artefact, the user can include it in the model itself, by adding attributes to the model's elements.

When analysing a requirements model, the user can apply the proposed visualization strategies in order to attain a clear comparison of the attributes pertaining to different elements. For instance, one may want to identify the elements that have higher volatility, or to compare the cost-benefit-risk of different elements, through ranged value visualization. Additionally, the user can apply the binary visualization functionality in order to quickly grasp the requirements that are scheduled for a particular release. By visually checking the metadata information, the user may identify adjustments to be performed, then proceeding to make such adjustments and re-applying the visualizations.

The user interface of the piStar tool has three sections: top, left, and right (Figure 9). The left section is the properties panel, which displays the properties and contextual actions of the selected element of the model. This panel is used to manually add new properties/attributes (e.g., cost, risk, and priority) to individual elements, as well as to edit their values. The right section is the modelling canvas, on which the model itself is created and modified. On the top section of the screen there is the main menu area, with five options: File, Add, Options, Help and Visualization plugin. The last menu provides access to the visualization functionalities:

Manage attributes provides a straightforward way to add attributes to every element in the model.

Visualize range presents the options to visualize ranged values, as shown in Figure 8-a: colouring (“Change colour based on...”), resize (“Change size based on...”), and icons (“Show attribute X”). For each visualization option the user chooses

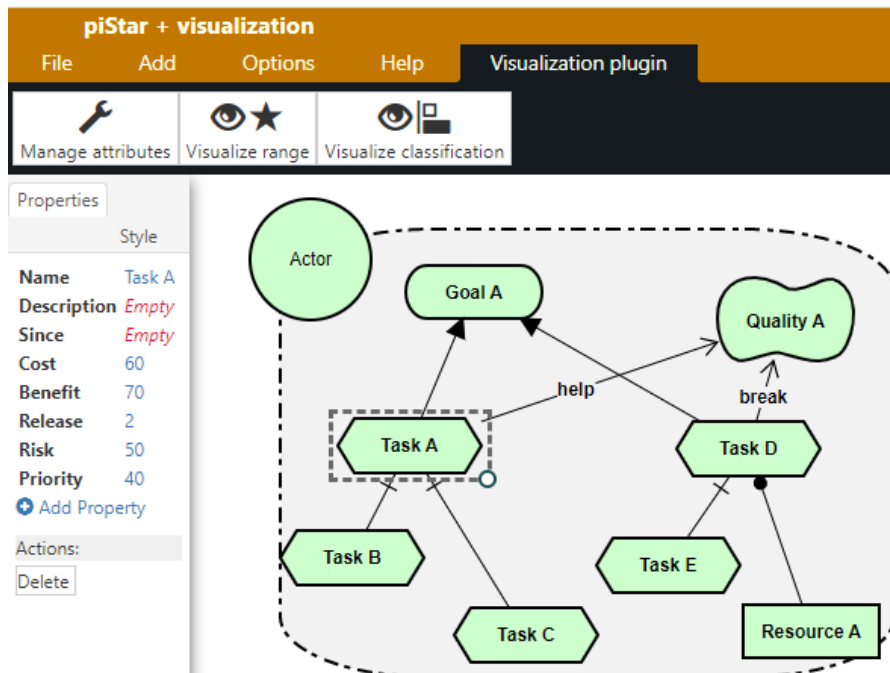


Figure 9. Main screen of the piStar tool with the visualization plugin. Top: Main menu; Left: Properties Panel; Right: Modelling canvas.

on which attribute it will be based. Up to three icons can be displayed, each with the options of presenting values, colouring and resize on the icon itself.

Visualize classification: allows highlighting the elements according to a certain criteria such as, for instance, whether they pertain to a given release (Figure 8-b). The user may choose how to highlight them: through line decoration, colouring, resize, or a combination thereof.

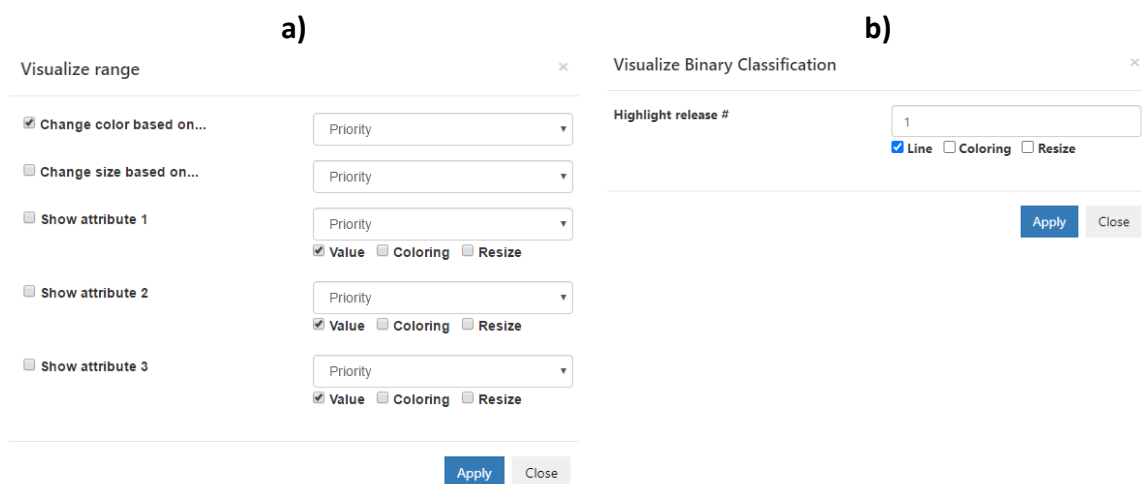


Figure 8. Visualization options: a) range visualization options; b) binary classification visualization options.

4.1. Architecture and design

The tool adheres to the Model-View-Controller (MVC) architectural style. In particular, the *i** model has two views: the graphical view, created as an SVG (Scalable Vector Graphics) image; and a tabular view, to represent the attributes of each individual requirement. The user interface is also mostly developed as models and views, except for the overall layout of the tool and some static elements, which were written with plain HTML (HyperText Markup Language).

The *i** diagrams can be exported as images for use with other tools. Two options are available: export as SVG and export as PNG (Portable Network Graphics). The former is the highest fidelity option, as it is a vector file format. Nevertheless, since some tools do not support this file format (e.g., older versions of the Microsoft Office suite), we also developed the option to export as high-resolution PNG images. The model itself can be saved locally as JSON (JavaScript Object Notation) object. Figure 10 shows an excerpt of a goal model with its corresponding JSON object, where an actor named “Medi@” has a task named “Internet Shop Managed”. This task has cost 70, benefit 30, and risk 15.

The following libraries were used in this project: JointJS, to handle the edition of visual diagrams; jQuery and x-editable for the user interface; and Backbone for implementing the MVC style. The Bootstrap framework was used to provide a pleasant look-and-feel to the user interface.

4.1. Evaluation

The execution time for the application of a visualization strategy is a relevant acceptance factor for this proposal. This is the case since, if the person that wants to analyse the visualization needs to wait too long to actually see the visualization, he/she may get frustrated and give up on it. Thus, we performed an evaluation of the execution time for applying the most processor-intensive visualization mechanisms: colouring and resizing.

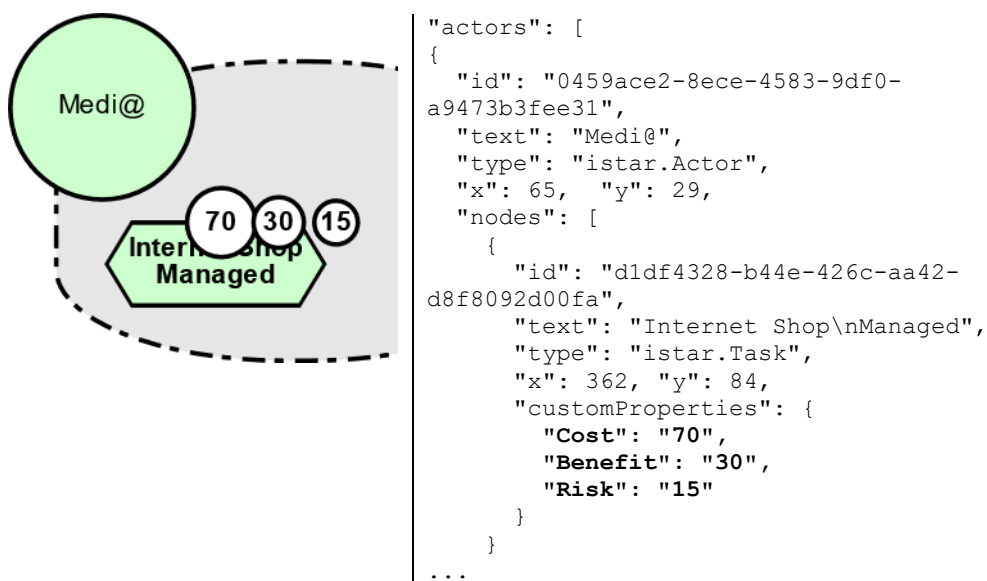


Figure 10. JSON object representing an excerpt of an *i** model

The largest model we were able to find in public repositories was presented in a collection of goal modelling showcases [Maiden et al. 2011]; it contains approximately 493 elements. In a study that analysed the complexity of goal models, the larger example that was found contains approximately 200 elements [Gralha, Araújo and Goulão 2015]. Considering a safety margin of 100 percent, in order to perform this experiment, we created random models with one thousand elements.

We developed a procedure to create a model with 1000 elements, where the kind of each element (goal, quality, task, and resource) is assigned randomly using JavaScript's default random function. Each element also has an associated benefit value, ranging from 1 to 100. Each experiment was repeated ten times on a fresh browser tab, to reduce interference from factors such as operating system tasks and browser's garbage collector. Additionally, a new model was created for each repetition of the experiment, to reduce the number of confounding factors.

The average execution time for colouring (i.e., changing the colour of every element according to the value of one of its attributes) was 7269.88 milliseconds, with a standard deviation of 292.84. This time includes not only the processing time (which decides which colour to apply), but also the time required for the browser to actually update the colour of each element.

Resizing (i.e., changing the size of each element according to the value of one of its attributes) presented an average execution time of 2466.72 milliseconds, with a standard deviation of 134.24.

Figure 11 shows the execution time for each repetition of the experiment. The experiment was run on a 64-bits machine, with an Intel i5 2.2GHz processor, running Microsoft Windows 10, using the Google Chrome web-browser.

Even though experimentation with real users is required to assess if these execution times are satisfactory, we believe that taking 7269.88 milliseconds for exceptionally large models will not compromise the users' workflow. Furthermore, subsequent colouring and resizing that would be expected on iterative scenarios are expected to take less than one second, since the browser skips the re-colouring of any given element if the new colour is the same as the old colour.

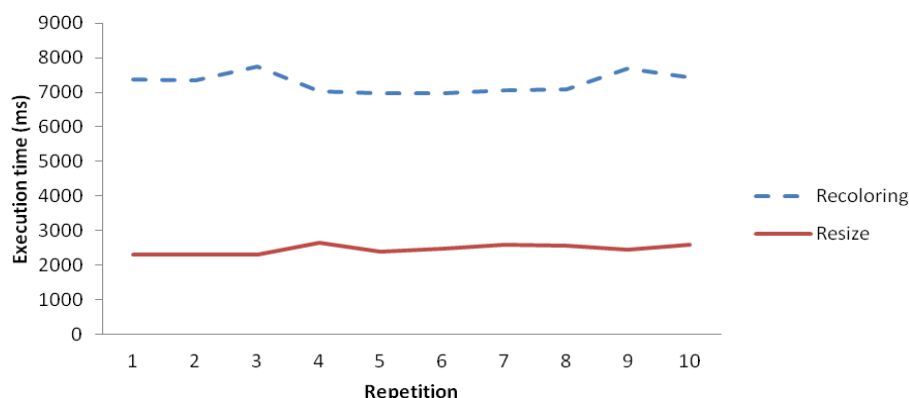


Figure 11. Execution time for each repetition of the experiment

5. Related Work

In this section we discuss related work, focusing on goal-based approaches for requirements prioritization or release planning. The work of Liaskos et al. [Liaskos, Jalman and Aranda 2012] presents a proposal to use the Analytic hierarchy process (AHP) with i^* models, aiming to obtain quantitative indicators of softgoal satisfaction. Whereas Liaskos work adopts a prioritization technique to elicit new information, our proposal is different since it aims to visually represent prioritization information.

Horkoff and Yu [Horkoff and Yu 2010] propose an approach to select alternatives on i^* models, based on desired levels of satisfaction of softgoals. This is achieved with an algorithm that, when unable to find an optimal solution, asks the user for additional information. Horkoff and Yu's work focuses on selection, more specifically on decision making, not on prioritization. Both [Liaskos, Jalman and Aranda 2012] and [Horkoff and Yu 2010] can be used together with our proposal, as follows. Liaskos et al. [Liaskos, Jalman and Aranda 2012] enables the elicitation of precise values for contribution links, which can aid stakeholders when prioritizing the elements of the i^* model. On the other hand, stakeholders can use the visualization strategies here proposed in order to inform the decision-making of Horkoff and Yu [Horkoff and Yu 2010].

The work of Kassab [Kassab 2013] demonstrates the application of a decision-making technique, AHP, along with the NFR Framework [Chung et al. 2000]. It aims at supporting trade-off analysis of the operationalizations of non-functional requirements, in the context of quantitative approaches. Such a technique can provide the raw data that will be visualized with our proposal.

The purpose of Regnell et al. [Regnell and Kuchcinski 2013] is to provide support for combinatorial optimization in the requirements engineering domain, adopting reqT [Regnell 2013] as the base requirements language. Using constraint satisfaction programming (CSP) and an automated solver, that proposal allows calculating values, to find solutions to constraints, and to perform queries on requirements models. It provides examples of requirements prioritization, release planning, and product line modelling. Whereas Regnell's approach focus on processing support for automatic evaluation, our proposal focuses on visual support for manual analysis.

Aydemir et al. [Aydemir et al. 2018] propose a goal-based approach for addressing the next release problem – that is, selecting which requirements to implement next, from a set of candidate requirements. They extend traditional goal trees with a set of inter-dependency constraints, such as REQUIRES (R1 requires R2 to function) and TEMPORAL (R1 needs to be implemented before R2). Such extension can be helpful for the analysis of release planning, representing a promising venue for future work.

6. Conclusion

In this paper we present a set of strategies to support the visualization of prioritization-related information and the use of said information in the context of release planning. Combining the expressiveness of i^* models with the proposed strategies for visualizing prioritization and release planning attributes, the proposal potentially provides rich

capabilities for grasping, analysing, reviewing, and updating prioritization information, furthermore facilitating the decision-making process of release planning.

The approach is supported by a web-based tool, which has been evaluated for performance. The customization of the application of visualization strategies is a strong point of the tool, allowing for the use of different prioritization criteria, prioritization approaches, and visualization preferences. However, the large number of options may be daunting for some users. Thus, as future work we expect to perform empirical studies and identify the most effective combinations of visualization strategies, analysing their impact on the cognitive load of viewers of the model.

We also plan to analyse the use of additional inter-dependencies as discussed in [Aydemir et al. 2018], in order to provide even more information for the viewers of the models. Lastly, we plan to expand the visualization strategies beyond the scope of requirements prioritization and release planning – e.g., for tracking software development in terms of elements to do/doing/done.

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