Abstract—Traceability helps explaining the execution and evolution of software and systems. It is a key input in many engineering tasks such as program understanding, maintenance and debugging. Several metamodels to facilitate the representation of traces and links among related artefacts have been proposed. Nevertheless, we claim they lack the mechanisms to express important traceability aspects such as the quality of traces, their gradual decay, and the evidences supporting them. This affects the benefits traceability can bring to the above-mentioned tasks. This paper presents a more expressive traceability metamodel, covering all the missing dimensions in a single, but extensible and modular, design. It characterizes trace quality to consider traces as salient artifacts in system and software development and maintenance. This modularity facilitates the integration of our solution in other modeling languages or its partial adoption when only some specific traceability aspects are needed. Its extensibility facilitates its customization (e.g., in terms of the types of links and artefacts) to better cover specific domains.

Index Terms—System and Software Engineering, Model-Driven Development, Traceability, Metamodeling

I. INTRODUCTION

Traceability is the ability to trace different artefacts of a system (of systems). It is defined in the IEEE Standard Glossary of Software Engineering Terminology [25] as the degree to which a relationship can be established between two or more products of the development process, especially products having a predecessor–successor or master–subordinate relationship to one another.

The need for traceability has always been a recurrent aspect of systems and software development. Across the years, there has been a continuous interest in developing techniques to facilitate the representation and analysis of traces and links between related artefacts. It helps explaining their execution and evolution as traces offer a different perspective on a system, arbitrary and customizable, where the relationships between elements is the most salient artefact. Traceability raises awareness on specific purposes or goals [12] and has been proven useful in a diverse number of software engineering challenges [21]. It is inherent to any software maintenance effort such as change impact prediction [24], debugging [29], [1], feature location [13], [34] or certification [35] among many others [27]. In the trenches of DevOps, traceability maintenance is essential to cope with the continuous artefact changes for successful system and software project management [39].

In the literature, the importance of traceability is reflected in the production of many metamodels targeting the specification of traceability aspects. Many of them focus on specific aspects or domains where the traceability mechanism is applied [3], [47], [46]. There is even work dedicated to the engineering of metamodels for traceability that offers a language dedicated to defining traceability metamodels [14]. Metamodels flourish, but their knowledge and contributions remain scattered among systems and software engineering research fields. Overall, we are still missing a generic metamodel for traceability that covers not only the representation of artefacts, traces and links between them but also quality aspects that can be used to interpret the relevance and integrity of traces.

AI can be a mechanism to infer new traces among sets of artefacts [6], [21], which will need to come hand in hand with the proper explainability support, specially considering the non-deterministic nature of information retrieval algorithms. This non-deterministic nature draws in a significant degree of uncertainty about the results such algorithms may yield. Traces automatically identified show variable confidence. In the same fashion, versatile artefacts see their cohesion with traces decreases with time. This dimension should be considered as a core concern to traceability.

In this sense, the contribution of this paper is the definition of Tracea: a generic and extensible traceability metamodel integrating quality concerns (e.g. decay, confidence, and explainability) in the definition of traces. The design of the metamodel favours also its adaptation to specific application domains and model-driven tool-chains to open the door to a new generation of techniques (e.g., for impact analysis) that could benefit from our more expressive metamodel.

The rest of this paper is structured as follows. Section II introduces the state of the art through a comparison between a selected set of approaches addressing the modeling of traceability. Section III presents the limitations of current solutions through their main quality concerns. In Section IV, we show and depict our metamodel Tracea and display an illustrative example. We discuss about the integration of Tracea and more generally of traceability modeling into existing tooling in Section VI before we conclude in Section VII.

II. STATE OF THE ART

In a previous work, we have identified over 80 approaches aimed at modeling traces and tracing activities [5]. We describe in this section a selection of the most representative
publications and then summarize their main limitations as the key challenges our solution will aim to overcome.

Most works on modelling traceability come, historically, from the requirement modelling community. Traces are seen as "links" from requirements to their (sub)components, and to their design and/or implementation artefacts [44], [22]. Specially relevant is the work by Goknil et al. [19] that includes a metamodel for traces, mechanisms for consistency checking and inferencing, and tooling for change impact analysis.

In the model-driven engineering (MDE) community, research work can be classified into proposals focusing on modeling the heterogeneity of artefacts – with numerous contributions aiming at linking text artefacts to design models [41], or addressing the entanglement of (non functional) requirements [48] – and proposals adapting traceability to specific application areas such as the automotive and robotic industry [15], [40], [42]. In this group, we see several publications that include custom metamodels built ad hoc to solve specific model transformation issues [30], [28].

Other modeling approaches are aimed at establishing an automated trace generation process, e.g., for requirement traceability. For example, Spanoudakis et al. identify rules to generate links automatically [43]. We see these cases as sidesteps from traceability modeling since the works aim at generating traces (that need to be modelled) rather than modelling traces (that need to be generated). As a consequence, the presented metamodels are specific to the types of models of the generation (e.g., BPMN models [38], or data warehouse models [33]). Natural language is also often used in this type of automated processes. In this case, approaches target the extraction of semantics (or meaning) from textual requirements. These publications model text blocks with their dependencies and the dependencies to specific third party artefacts (e.g., for MDE: [41], for AADL [45], for agile user stories [10]).

Recently, researchers attempted more generic approaches to traceability, closer to our own goal. Building on previous knowledge in specific domains, authors describe their attempts to synthesize traceability requirements. For example, Azavedo et al. [4] created a metamodel with explicit (57) relationship types and (12) different kinds of artefacts based on an arbitrary separation of software development tasks (e.g., Implementation, Verification, Modification, Homologation). On the other hand, Heisig et al. present a modeling approach to traceability that includes both a basic metamodel with a plugin mechanism (using XText) that allows users to define their specific representations for links and artefacts [23].

While these latter approaches do represent an advance in the generalizability and adaptability of traceability metamodels, our approach offers a higher granularity and decomposition while integrating several quality concerns (decay, confidence, and explainability). Table I summarizes existing works regarding these core traceability aspects. As shown in the table, most publications consider a single trace level, which limits the complexity and diversity of problems where traceability can be applied. There is also a significant lack of consideration for quality aspects. Consistency is merely mentioned and confidence is strictly forgotten – none of the selected approaches mention it. Explainability is reported in a few cases but remains scarce and no common appreciation has emerged yet.

III. TRACEABILITY REQUIREMENTS

Following up on the state of the art analysis, this section details the traceability requirements we believe are needed in order to have a complete traceability modeling solution, able to be used in a variety of scenarios, including industrial applications. Next section describes our proposed metamodel and how its different components satisfy these requirements.

A. Adaptability & Configurability

Reusability of a traceability solution is key for its industrial adoption. When traces are seen as useful only to conform a very specific requirement (e.g. software certification deadline), enterprises have shown that it is easier or cheaper to execute it as a manual and ad hoc process [12].

We aim for a metamodel that is configurable and adaptable to maximize its reusability in a number of application scenarios thus favouring its adoption by companies. For example, a specific certification paragraph might be better suited (i.e., more precise) for the user than the entire certification document containing this paragraph. Or, if the purpose is to trace the impact of changes in a model on the source code, does the user want to know about the occurrence or the location of a change? Does the user want to find the right file or the right class, the right method, or the right package? In other words, what kind of artefacts of the software product is of interest (e.g., design models, source code) and to which level of granularity?

High-level types for artefacts as well as peculiar level of granularity must be adequately designed in the tracing solution.

1) Configurable tracing: A trace is commonly expressed as the combination of atomic trace links representing direct connections between a number of artefacts. For example, a certification document (e.g., ISO-26262 [26]) is "linked" or "related" to a set of design documents, or models, themselves being used for (or "relating to") the generation of source code or other related artefacts such as behavioral models [31]. Depending on authors intentions and problem constraints, they define traces with a single or multiple sources and end with one or many targets.

There is little attention put on more complex tracing purposes such as the complete sequence from certification specifications to source code implementation, or long reach tracing ability involving sequences of artefacts or decisions in chain.

We believe a traceability metamodel must come with several levels of granularity to enable users express traceability relationships either at a coarse-grained or a fine-grained level depending on their needs. Moreover, if defined at a fine-grained level, the model should be able to use that information
### B. Consistency

One of the main arguments against investing in automated traceability support remains the cost of maintaining traces up-to-date [12]. Software systems evolve and endure maintenance bug fixing and patches that can potentially modify their constituent elements at every level. Even their architecture changes through time to cope with increasing scalability needs, to comply with new privacy regulations, or to add or modify the panel of features offered to the different kind of "users" of the system. Tracing is no alien to this phenomenon and the cost to maintain traces consistent with the system increases hand in hand with the system volatility.

There is no consensus on the means to ensure that traces remain consistent to the system. Yet, the naïve method that consist in rebuilding the entire graph of traces each time from scratch does not scale [42]. Gervasi et al. exploit the information contained in previously defined traces, in order to facilitate the creation and ongoing maintenance of traces, as the requirements evolve [18]. Seibel et al. have shown that the MDE paradigm offers auspicious horizon to the maintenance of traces [42]. They execute rules in order to maintain a set of links representative to the trace types predefined beforehand. Authors extend the concept of timestamp to consider context changes and thus to reflect better the system volatility.

We agree that traceability metamodels should be able to represent temporal information [8], not only for the traces but for all the traced elements so that we can compare them and evaluate the potential decay the traces may suffer.

#### C. Confidence

Decay is not the only factor that can affect our confidence on the consistency and relevance of a trace. The execution of automated processes to identify traces raises uncertainty about the actual existence of the results they yield. Learning techniques, using deep learning algorithms such as in [21], offer to bridge the cognitive gap among artefacts of different nature but accuracy is never perfect. There lies an open topic at the intersection between "traditional" and "AI-enabled" software practices. Systems with AI-enabled components [generally probabilistic] can have a high margin of error due to the uncertainty that often follows predictive algorithms [37].

Taking account of the non deterministic nature of AI modules is a key factor for AI-enabled software of quality. Moreover, even a manual trace identification process can have some uncertainty as designers may not be completely sure about the real relationship between components that may have been created a long time ago.

Therefore, we need to be able to express in a traceability model the confidence we have on the traces. Where to draw the line between a useful trace based on this confidence level depends on the envisioned application. There is a trade-off to evaluate between the level of confidence and the level of criticality of the project. If the purpose is to evaluate a requirement change impact on the source code, traces with a low confidence level may trigger false positive and generate some additional work but still be reasonably useful. If traces

<table>
<thead>
<tr>
<th>Approaches/Quality</th>
<th>Adaptability</th>
<th>Granularity</th>
<th>Consistency</th>
<th>Confidence</th>
<th>Explainability</th>
</tr>
</thead>
<tbody>
<tr>
<td>Goknil et al. [19]</td>
<td>Generic types</td>
<td>1-step links</td>
<td>–</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Taromirad et al. [44]</td>
<td>Fixed types</td>
<td>1-step links</td>
<td>–</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Haidrar et al. [22]</td>
<td>Fixed types</td>
<td>1-step links</td>
<td>Timelessness</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Sanniet et al. [41]</td>
<td>Specific types</td>
<td>1-step links</td>
<td>–</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Dubois et al. [15]</td>
<td>Specific types</td>
<td>Multi steps</td>
<td>–</td>
<td>–</td>
<td>Evidences</td>
</tr>
<tr>
<td>Sanchez et al. [40]</td>
<td>Specific types</td>
<td>Multi steps</td>
<td>–</td>
<td>–</td>
<td>Evidences</td>
</tr>
<tr>
<td>Yrjonen et al. [48]</td>
<td>Specific types</td>
<td>Multi steps</td>
<td>Timelessness</td>
<td>–</td>
<td>Evidences</td>
</tr>
<tr>
<td>Jimenez et al. [28]</td>
<td>Specific types</td>
<td>Multi steps</td>
<td>(not applicable)</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Wang et al. [45]</td>
<td>Fixed types</td>
<td>1-step links</td>
<td>–</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Camiel et al. [10]</td>
<td>Fixed types</td>
<td>Multi steps</td>
<td>–</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Spanoudakis et al. [43]</td>
<td>Generic types</td>
<td>–</td>
<td>–</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Pavalkis et al. [38]</td>
<td>Specific types</td>
<td>1-step links</td>
<td>–</td>
<td>–</td>
<td>Agent</td>
</tr>
<tr>
<td>Maté et al. [33]</td>
<td>Specific types</td>
<td>1-step links</td>
<td>–</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Azvedo et al. [4]</td>
<td>Generic types</td>
<td>Multi steps</td>
<td>Timeliness</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Heisig et al. [23]</td>
<td>Generic types</td>
<td>Compositional</td>
<td>Context sensitive</td>
<td>–</td>
<td>–</td>
</tr>
</tbody>
</table>

**TABLE I**

**OCCURRENCES OF THE MAIN PROPERTIES FOR MODELING TRACEABILITY.**
are used as part of a security certification, a high confidence level is a must to obtain valid results. The propagation of uncertainty is an open topic that authors attempt to address with a mix of boolean logic and Gaussian statistics [7]. That should be applied to trace uncertainty as well.

D. Explainability

Explainability is more and more important in any software system due to increasing transparency, and ethical and regulatory concerns. Users do not only require that the answer of the system is the one expected, they need to know how did the system proceed to yield such answer. AI-enabled systems need to be able to register this nature as well as the details of the identification process.

For example, programmers often use mnemonics for identifiers that help associate code with high-level concepts in the requirements and vice-versa [2]. If traces, used for quality audit, have been identified thanks to a rule-based approach exploiting this mnemonics, this information needs to be part of the traceability model. Those traces could be later used as evidences to automatically check for potential mismatches or coverage analysis of requirements not supported in the implementation.

IV. Tracea Metamodel

Based on the previous analysis, we present in this section a new traceability metamodel, called Tracea. We aim not to present yet another traceability metamodel but one that learns from and supersedes previous proposals in order to provide a more complete metamodel that also responds to the requirements made explicit in Section III. In the course of producing this metamodel, we have tried to be as objective and inclusive as possible. To this extent we use existing knowledge on adaptable and consistent traceability which we augment with quality aspects (confidence and explainability) found missing in the literature of the field. To foster legibility, we have added a simple but complete example of the use of our metamodel in the following section.

A. Adaptable and configurable traces

In Tracea, we start from the common core trace representation found at the intersection of several of the existing metamodels and expand and refine it with a more configurable structure that allows one to define the proper level of granularity.

1) Fine grain tracing structure: The excerpt in Fig. 1 describes the composition scheme of links into a forest-like structure. Its atomic elements are TraceLinks. A trace link refers to a source and a target ArtefactFragment (see below). It may be a leaf – which means it has no successor links; or a node – which means the trace does not end with this link. A TraceLink is a composite of LeafTraceLink and NodeTraceLink. A Trace starts from a set of trace links "firstLevel" that connect to their respective trees. The set of derived TraceLinks from the transitive closure of firstLevel is contained in the reference traceLinks. In the same manner, the set of source and target of traces (see blue references in Fig. 1) is derived as well (see Listing 1). Trace and TraceLink are subtypes of TracingElement with a unique identifier (name), a timestamp to address consistency issues (see below), and one or more Agent related.

```java
context Trace inv firstLevels:
  self.traceLinks
-includeAll(self.firstLevels)
context Trace inv sources:
  self.firstLevels
-includeAll(self.first.sources)
context Trace inv targets:
  self.traceLinks
-includeAll(self.targets)
context Trace inv sourceArtefacts:
  self.firstLevels
-includeAll(self.sourceArtefacts.fragments)
context Trace inv targetArtefacts:
  self.traceLinks
-includeAll(self.targetArtefacts.fragments)
```

Listing 1. OCL constraints for derived references

2) Adaptable artefacts and relationships: Existing traceability works differ a lot in the kind of artefacts they target. A unified ontology of traced artefacts has yet to emerge but a common approach is to distinguish between the nature of artefacts. I.e., from text intensive (e.g., requirements, certification) to structure intensive (e.g., source code, test cases). In Tracea, we specialize Artefact with TextArtefact, ModelArtefact, CodeArtefact, and TestArtefact. The list is not exhaustive. These high-level types support the user in defining her own (sub)types. They are anchors to refine artefacts and their fragments at an adequate level of granularity.

To freely adjust the granularity of the artefacts, Tracea suggests the fragmentation of Artefact. An ArtefactFragment defines a part of an artefact that is of interest (e.g., a method in a class; a section in a text document). Tracea implements a high-level separation for artefacts and relationships. This enables the separate customization of artefact types and the semantic relationships between them. The typing of relationships is versatile and we distinguish two kinds depending on the domain they apply to: DomainType and EngineeringType. In the former, the semantic of the final user (i.e., its domain of application) is targeted. In the latter, the concepts used by engineers or modelers are targeted (the domain of engineering). Same for the artefact types, relationship types are also expected to be customized [36] when needed.
Fig. 1 shows an excerpt of the Tracea metamodel that focuses on aspects related to the adaptability of our approach.

B. Confidence of trace links

In Tracea, the confidence of a trace (TraceConfidence) or of a trace link (LinkConfidence) is an operation returning a real number value representing the level (from 0 to 1) at which the trace is certain to exist in the system. It is a statement about the relevance of a trace. A trace and its links are made by Agents. As illustrated in Fig. 3, an agent can be human (e.g., when traces are elicited manually), or an agent can be a machine (e.g., when an algorithm identifies the trace automatically).

Another concern lies in the recording of trace evolution. The trace creation should be recorded, with the successive changes that affect it, for evolution analysis. Integrity measures respective to evolution events (e.g., creation, modification) should be recorded as well to evaluate their evolution during a period of time. We implement the occurrence of these events with the definition of attributes of tracing elements.

C. Trace Consistency

To address the issue of gradual decay, tracing elements must be considered alike with other software artefacts. Their evolution must be scrupulously and synchronously monitored. To be able to represent (and later reason) on the potential decay, we add timestamp attributes to TracingElement (see Fig. 3), de facto transforming all metaclasses inheriting from it in temporal elements. We can then use these timestamps to compare the age of a trace with the age of the elements traced by it and, if needed, update the confidence we have in that trace, or trace link, accordingly.

D. Explainability

Traces are a key element in many system and software engineering activities. Therefore, engineers may not want to just take them at face value but ask for explanations on how
and when the trace was created. Previous subsection covered the *when*, here we focus on the *how*.

The degree of confidence may be justified with evidences, if they exist, to explain the rationale behind the quantitative value. In case of links automatically identified, an evidence instance can record the information necessary to reproduce the identification process or at least to partially explain it.

More precisely, as can be seen in Fig. 3, an evidence refines into three sub-types: AnnotationEvidence contains a textual description; RuleEvidence contains a rule (or a set of rules) in a textual field, as well as the execution date; and AIEvidence contains attributes that will help reproduce the learning scenario, e.g., the kind of algorithm, a reference to the training set, the associated precision and recall of the algorithm, and others parameters. An Evidence explanation can also point to other supporting tracing elements. These elements testify, or illustrate the evidence and will be useful for later consistency check.

Every evidence is also optionally endorsed by a set of Human or Machine agents, which further helps in the explainability of the trace beyond the description and attribute values stored in the Evidence object itself.

V. ILLUSTRATIVE EXAMPLE

In this section, we introduce a simple illustrative example: tracing the impact of a change in the requirements onto its implementation in Java classes. We show through this example the customization of artefacts and relationships, the importance of a quality evaluation, and how we circumvent consequences of using AI-enabled modules for trace identification.

A. Project purpose

In this example, links are relating requirements to Java classes that undergo modifications. Java classes and requirements representation need to be customized in the tracing system, as well as the kinds of links that occur between them. If those links are identified automatically (e.g., thanks to an AI identification technique), they bear a confidence level related to the accuracy of the approach used (e.g., a combination of precision and recall). Evidences backing up this quantification (e.g., the kind of algorithm, the dataset used) can also be stored for later consultation. With this information recorded, Tracea enables users to answer questions such as "whose classes have been impacted by a specific change in the requirements?" Or more precisely, "what Java classes are impacted by this change in the requirements with a minimum confidence of 80%?"

B. Customization of artefacts and relationships

In this example, a trace aligns two kinds of artefacts: Requirement specification and Source class.

These artefacts are too complex to be used at a coarse level of granularity. Java classes may comprise hundreds (or even thousands) of lines of code, requirement specification documents contain hundreds of sections. To address this size issue, a source Artefact (e.g., a class) is decomposed into smaller part (such as methods). In the same manner, specification documents are decomposed into sections. Listing 2 shows an excerpt of our textual concrete syntax applied to this example where we can see the fragmentation of artefacts. The structure of the traced system is first described with artefacts and fragments sections. For legibility concern, the only kinds of relationships in that example are Implement, i.e., a source
class implements a requirement section, and Redirect, i.e., a login reject redirects to a different method.

```java
artefacts {
  Requirement r_01 {fragments {sAuth, sLogout}},
  Source Login.java {fragments {mLogin, mLogError, mLogout}},
}
fragments {
  RequirementSection sAuth { }, sLogout { },
  Method mLogin { }, mLogError { }, mLogout { },
}
relationshiptypes {
  EngineeringType Implements (),
  DomainType Redirects ()
}
agents {
  HumanAgent 5e8a51e4,
  MachineAgent Rd150u5rd
}
```

Listing 2. Artefacts, Fragments, Relationships, and Agents declaration

The concrete traces are recorded as illustrated in Listing 3. A Trace is identifiable by its name and contains trace links whose composition is described through successors. This example is the minimalist expression of a trace. Each and every element is susceptible to refer to an Agent that indicates who is responsible for the edification of the trace (and what is the nature of that “who”). In our case, link_01 and link_03 have agent referees.

As can be seen in Table I, five approaches allow the customization of artefacts and links [4], [19], [23], [30], [43]. Yet, these latter do not offer any mean to record whose agent is responsible for identification. We offer means for both.

```java
Trace ChangeImpact {
  tracelinks {
    NodeTraceLink link01 {
      source sAuth
      target mLogin
      successors (link02)
      relationshiptype Implements
      agents 5e8a51e4
      confidence cSure
      evolution ((20210621-0954),{}),
    },
    LeafTraceLink link02 {
      source mLogin
      target mLogError
      relationshiptype Redirects
      confidence cSure
      evolution ((20210621-0954),{}),
    },
    LeafTraceLink link03 {
      source sLogout
      target mLogout
      successors ()
      relationshiptype Implements
      agents Rd150u5rd
      confidence cSure
      evolution ((20210621-1521),{}),
  }
}
```

Listing 3. Trace instance

C. Explainability for AI-enabled traceability

You are not forced to create the above traces yourself, there exists automated evaluation techniques for change impact that predict which classes are most likely to change. Tracea also supports this scenario in which a change in the requirements links to potentially impacted classes, and to actually modified classes. This distinction shows a distinction in nature of the links themselves. The former is more inclined to suffer a low level of confidence than the automatized latter. In our case, Listing 3 shows that links link_01 and link_02 have been manually identified, and thus their confidence is 1.0 whereas Link_03 has been automatically suggested and boasts a confidence of 0.8. This level of confidence relies on evidences about the algorithm employed, its parameterization, and its training setting. As can be seen in Listing 4, a confidence is related to a Trace or a TraceLink and a set of Evidences.

```
confidences {
  Confidence c01 {
    value 0.8
    evidence (Evidence_link03)
  },
  Confidence cSure {
    value 1.0
    evidence ()
  },
}
evidences {
  AIEvidence Evidence_link03 {
    algorithmUsed "AI4All"
    parameters ["platform:\/resource/train/pos_202012"]
    executionDate "20201207-123536"
    trainingResults .8 .7
    impactedElements {*link02*, mLogin, otherMethod}
  },
}
```

Listing 4. Confidence and evidences

D. Analysis capabilities

Once the information is stored in the Tracea system, this opens the door to a number of analysis queries. An option to answer the aforementioned question about the identification of Java classes impacted by a change in the requirements with a minimum confidence of 80% is to implement it with OMG’s Object Constraint Language (OCL). Listing 5 shows an excerpt that captures the targets (Source class) of a trace if the confidence in the links that binds them to their source (Requirement) is greater than 0.8.

\[^1\text{omg.org/spec/OCL}\]
context Trace inv:
  self.traceLinks
  -select(tl : TraceLink | tl.confidence >= 0.8)
  .target

Listing 5. OCL implementation of confidence check

Other attributes can be used to further restrict the resulting set of classes. E.g., `evolution` timestamps can be compared to only consider recently active source files. "evolution" keyword in Tracea's syntax refers to respectively creation, modification and deletion time of links. Listing 6 illustrates a case where the user only wants to capture elements created before 2021.

Analysis is a specific case of traceability where traces are used to answer a specific question. Here we propose the integration of latter analysis factors in the deployment of tracing (e.g., confidence).

code

context Trace inv:
  self.traceLinks
  -select(tl : TraceLink | tl.confidence >= 0.8 && isBefore(tl.creationDate, "20210101") }
  .target

Listing 6. Confidence check with evolution constraint

VI. INTEGRATION AND TOOL SUPPORT

Beyond this option, we have integrated Tracea on top of An Xtext-based\(^2\) definition of our metamodel is available on Git\(^3\). As concrete syntax, we are using the JSON textual syntax shown in the examples above and illustrated in Fig. 5.

Capra\(^4\) to enable ubiquitous quality traceability for EMF elements. Eclipse Capra is a traceability management tool offering some interesting features to edit and visualize traces, including traceability matrices and graph visualisations. The customization language is based on Xcore\(^5\) and we added Tracea concepts as an extension of Capra to integrate our trace quality definition. The source code is available on Git\(^6\). This integration is an actionable example of the integration of quality attributes into an existing tracing software system. Thanks to it you can benefit from the advanced metamodeling concepts in Tracea while also enjoying Capra's visualization capabilities. As an example, with Tracea, Capra allows the filtering of links that satisfy a confidence threshold. In graphical representation, the confidence is shown together with the name or type of the links. In matrix-based representation, a color map is used to show the links that satisfy the threshold (see Fig. 4).

In both cases, the designer can use Tracea as a standalone tool or add it as a new component to any model-driven pipeline. Capra is especially relevant to exploit the most of the Eclipse EMF ecosystem.

As, in complex scenarios, traces can come from different systems (using different languages or even third-party APIs), it is useful to keep Tracea as an external language that you can adapt to the changing needs of your application scenario and the types of artefacts you need to trace \([11], [32]\). But as a trade-off, this forces designers to learn and add to their toolset a new language. An alternative option is to define Tracea as a kind of internal DSL, embedded in a more general modeling language like SysML or UML using the extension capabilities offered by them, e.g. UML profiles. We are currently working on the integration of Tracea into SysMLv2 to allow a transverse linkage of elements mixing both existing and custom relationship types.

VII. CONCLUSION

Traceability research is scattered among different subfields of system and software engineering. This results in diverse but partial solutions to represent traceability information. We have presented a complete traceability metamodel that aims to cover all its aspects, including the quality and uncertainty of specific traces, their decay, or the evidences that support them. This information is needed to make fully informed decisions based on trace data. Our proposal has also been designed with modularity and extensibility principles in mind to facilitate its adoption in a large variety of domains. We believe it should help in improving a number of traceability-based algorithms (e.g., for change impact analysis) that could now also take into account these additional traceability dimensions.

As further work we want to continue advancing on these latter aspects, mainly proposing extensions to general modeling languages (like SysML or UML) that integrate our traceability metamodel. Moreover, we will explore the complementarity of AI and traceability. Regarding AI for traceability detection we plan to extend existing techniques to automatically infer traces to populate our metamodel considering the integrity and quality aspects of the inference process. Regarding traceability for AI, we plan to rely on our metamodel to offer better explainability support to the myriad of AI-based solutions for Software Engineering that right now mostly ignore this aspect \([9], [37]\).

acknowledgement

\(\text{\textsuperscript{2}}\)https://www.eclipse.org/Xtext/
\(\text{\textsuperscript{3}}\)https://github.com/ebatot/tracea-dsl
\(\text{\textsuperscript{4}}\)https://projects.eclipse.org/projects/modeling.capra
\(\text{\textsuperscript{5}}\)https://wiki.eclipse.org/Xcore
\(\text{\textsuperscript{6}}\)https://github.com/ebatot/tracea-capra
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