Crowd-Sourced API Documentation

Gias Uddin SWAT Lab gias.uddin@mail.mcgill.ca

Foutse Khomh SWAT Lab foutse.khomh@polymtl.ca

Chanchal K Roy Department of Computer Science Polytechnique Montréal, QC, Canada Polytechnique Montréal, QC, Canada University of Saskatchewan, SK, Canada croy@cs.usask.ca

Abstract—The learnability of an API suffers when the official documentation of the API can be incomplete. Developers discuss usage scenarios of APIs in the online developer forums. As such, by automatically mining such crowd-sourced documentation of APIs, we can address the shortcomings of API official documentation. We present a framework to automatically mine usage scenarios about APIs from online developer forums. Each usage scenario of an API consists a code example, a summary description, and the reactions (i.e., positive and negative opinions) of other developers towards the code example. We evaluate our API usage mining framework by producing a benchmark dataset. We observed a precision of 0.957 and a recall of 1.0 with the linking of a code example to an API mention in the forum.

Index Terms-API, Mining, Usage, Documentation.

I. INTRODUCTION

APIs (Application Programming Interfaces) offer interfaces to reusable software components. The learnability of an API depends on the availability and usefulness of learning resources of the API [1]. Unfortunately, despite developers' reliance on API official documentation as a major resource for learning and using APIs, the documentation can often be incomplete, incorrect, obsolete, and not usable [2].

The shortcomings in the official documentation led to the creation and popularity of online developer forums (e.g., Stack Overflow), where developers can ask and answer questions on how to address diverse development tasks that may involve APIs. A number of research efforts have focused on integrating information from the forum posts into API official documentation, such as the linking of API types in Javadocs to code examples in forum posts [3], the presentation of interesting textual contents from Stack Overflow about an API type in Javadocs [4] and so on. Unfortunately, the approaches can have the following shortcomings: 1) The traceability techniques cannot link a code example to an API mentioned in the post and about which code example is actually provided, 2) The techniques do not provide a textual description of what the code example does, and 3) The techniques do not offer any insights into the quality of the mined code examples.

We present a framework to automatically mine API usage scenarios from developer forum posts. Each scenario consists of four items: (1) A code example as discussed in a forum post, (2) An API about which the code example is provided to address a development task, (3) A natural language description of the code example to summarize what the code does, and (4) A list of reactions as positive or negative opinions from other developers towards the code example.

Our proposed API usage scenario mining framework works as follows. First, given as input a forum post that consists of a code example, we automatically link the code example to an API mentioned in the textual contents of the forum post (i.e., the code example is provided to discuss a use case of the API). Second, given as input a code example in a forum post that is linked to an API mention, we associate a summary description of the textual contents of the post where the code example is found. Third, we associate the positive and negative opinionated sentences as reactions towards the code example. In a benchmark-based study of 730 randomly sampled code examples from Stack Overflow, we observed that our algorithm to link a code example to an API mention in forum post showed a precision of 0.957 and a recall of 1.0. We compared the algorithm against two state of the art baselines. Our algorithms outperformed all the baselines.

II. RELATED WORK

The focus of this paper is to establish a framework to automatically mine usage scenarios about APIs from developer forums that can facilitate task-orientation documentation for APIs. We follow the concept of "minimal manual" which promotes task-centric documentation of manual. The format is proven to work better than the traditional API documentation [5]. We differ from the above work as follows: 1) We include comments as posted in a forum post as reactions to a code example in our usage scenarios. 2) We automatically mine such usage scenarios from online forum posts, thereby greatly reducing the time and complexity that may be required to produce those manually. Our algorithm to mine usage scenarios differ from the state of the art techniques to link code examples to APIs in forum posts [3], [6], because we detect an API as an API name as mentioned in the textual contents in the forum posts. Unlike the above work that consider each class of a software library as an API, we consider the entire library as an API. This design decision is based on our observation of how API names are actually mentioned and discussed in the forum posts. In contrast to [3] that needs the construction of an offline API database, our technique can rely on online API databases. Unlike [6], we do not rely on the analysis of client software code to infer usage patterns of an API. While such analysis can offer better accuracy than technique [3] using API databases, the approach is not feasible when such client software code may not be available (e.g., for a new API).



Fig. 1. The major components of our API usage scenario mining framework

III. THE FRAMEWORK

Our mining framework takes as input a forum post and outputs all the usage scenarios found in the post. Our framework consists of five major components (Figure 1):

C1. An **API database** to identify the API mentions. Our API database consists of open source and official Java APIs. An open-source API is identified by a name. An API consists of one or more modules. Each module can have one or more source code packages. Each package can have one or more code elements, such as classes, methods, etc.

C2. A suite of **Parsers** to preprocess the forum post contents. Given as input a forum post, we preprocess its content as follows: (1) We categorize the post content into three types: (a) *code snippets*; (b) *hyperlinks*; and (c) *natural language text* representing the rest of the content. (2) We detect individual sentences in the *natural language text*. (3) Following [7], we discard the following *invalid* code examples during our parsing: (a) Non-code snippets (e.g., XML, JSON extract), (b) Non-Java snippets (e.g., JavaScript). The rest of the code examples are considered as *valid*.

C3. A **Linker** to associate a code example to an API mention. Given as input a code example in a forum post, we associate it to an API mentioned in the post in two steps: a) We detect API mentions in the textual contents of forum posts following Uddin and Robillard [8]. Therefore, each API mention is a token (or a series of tokens) if it matches at least one API or module name. b) We associate a code example in a forum post to an API mention by learning how API elements in the code example may be connected to a candidate API in the mention candidate lists of the API mentions in the same post. We call this *proximity-based* learning, because we start with the API mentions that are closer to the code example.

C4. A **Generator** to produce a natural language summary description of a code example. Our algorithm to generate natural lanugage summary description of a code example takes as input all the textual contents of Stack Overflow thread where the code example is found and outputs a short textual summary description of the code example. Our algorithm works by first finding sentences where the API associated to a code example is mentioned both *explicitly* (e.g., by the API name) and *implicitly* (e.g., by a reference, such as a pronoun). We produce summary description only for code examples that are found in the answers to questions. This is based on the observation that such a code example is in more need to be understood within the context of a development task [3].

C5. An **Associator** to find reactions towards code examples. The inputs to the algorithm are all the comments towards the post where the code example is found. The output is a list of opinionated sentences that are related to the code example, e.g., the opinion refers to the API linked to the code example.

IV. EVALUATION

The effectiveness of a crowd-sourced API documentation technique relies on the correct linking of a code example to an API about which the code example is provided. Therefore, our initial investigation focused on the feasibility of our mining framework. We report using three performance measures: precision (P), recall(R), and F-measure (F1).

$$P=\frac{TP}{TP+FP},\ R=\frac{TP}{TP+FN},\ F1=2*\frac{P*R}{P+R}$$

TP = Nb. of true positives, and FN = Nb. false negatives.

We assess the performance of our algorithm to link a code example to an API mention using a benchmark. The benchmark consists of randomly selected 730 code examples from our entire dataset. We compare our algorithm against two baselines: 1) Baker [3], and 2) Search engine. We achieved a precision of 0.957 and a recall of 1.0 using our algorithm. Almost one-third of the misclassified associations happened due to the code example either being written in programming languages other than Java or the code example being invalid. The baseline Baker shows the best precision among all (0.973), but with the lowest recall (0.486). The Google search results show the lowest precision (0.389), confirming the assumption that Google is primarily a generally purpose search engine.

V. CONCLUSIONS

Developers discuss API usage scenarios in forum posts. Due to the plethora of APIs discussed in forum posts, it can be challenging to get a quick and informed insights from forums about APIs. We present a framework to automatically mine API usage scenarios from forums. To assist developers in their development task completion using the mined scenarios, we developed on online search and summarization engine for API usage scenarios. Our future work will focus on the further utilization of the mined scenarios to research improvement opportunities in API learning and documentation resources.

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