Reasoning about assessing and improving the candidate-seed quality of a generative aspect mining technique

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Abstract

We propose a new measure for assessing generative aspect mining techniques: the candidate-seed quality. We show the relevance of this measure and investigate how it can be improved for fan-in analysis, a generative mining technique that we have proposed in a previous work. The investigation results in a number of properties aimed at improving the quality of the candidate-seeds reported by fan-in analysis.

1. Introduction

Aspect mining is a research area aimed at developing techniques for automatic identification of crosscutting concerns in existing code. A number of approaches to aspect mining provide support for identifying and investigating the code pertaining to a crosscutting concern starting from a seed: a program element (method, interface or group of statements) that is part of the crosscutting concern implementation. We describe these approaches as explorative or query-based approaches.

A second category of aspect mining approaches focus on identifying crosscutting concern seeds by looking in the source code for symptoms of crosscutting behavior, like tangling and scattering. We group these techniques in the generative category.

The mining techniques, and the generative ones in particular, face serious challenges in describing, comparing and combining their results due to (1) the lack of a clear definition of the crosscutting concerns, and (2) their focus on generic symptoms (tangling and scattering) that can occur in various concerns, of different level of granularity. This poses further questions about how the results of a combination of aspect mining techniques can be assessed, how to show improvement of results, and what measures are relevant for assessing these results and a mining technique in general.

To address these problems, we have proposed a classification system of the crosscutting concerns based on sorts (atomic, generic crosscutting concerns) and a preliminary set of sorts [4, 5]. The crosscutting concern sorts are described by their specific symptoms (i.e., implementation idiom in a non-aspect-oriented language), and a (desired) aspect mechanism to refactor a sort’s concrete instances to an aspect-based solution. Sorts are relevant for aspect mining because they associate symptoms to generic concerns and are able to provide a focus for mining techniques and criteria for comparing their findings. Moreover, they are by definition atomic elements, and hence able to ensure a consistent granularity level for comparison.

In this work we further consider how the (generative) aspect mining techniques can be assessed, by proposing and discussing a new measure: the (candidate-)seed quality is an indicator of the human effort involved for analyzing the results of an aspect mining technique. We look in more detail at this measure and how it can be improved for a specific mining technique, namely fan-in analysis. Our investigation results in a number of properties to be considered for improving the (candidate-)seed quality for this technique.

2. Assessing generative mining techniques

The results of a generative aspect mining technique are candidate-seeds: program elements that might pertain to a crosscutting concern implementation, and which can be marked as either seeds or false positives by a human analyzer examining the output of the technique. The decision of the human analyzer is typically supported by a number of elements provided by the mining technique for reasoning about its output: grouping of results by some criteria (e.g., naming), (structural) relationships between properties describing the results (e.g., related method calls), etc.

For simplification, we will refer to candidate-seeds in the following sections simply as candidates. We will also use seed quality if we discuss about candidates that have been marked as seeds.

A serious limitation of the (generative) aspect mining techniques consists of lacking a clear definition of the crosscutting concerns and thus of those concerns they aim to identify. As a result, most of the techniques focus on generic symptoms of crosscutting behavior, like tangling and scattering. This makes the comparison and description of their results difficult:
• How is the reported candidate (program element) related to a crosscutting concern, and to which concern?
• How relevant is this relation for the mining technique: Is this an expected result or just an accidental match?
• Are the reported candidates relevant for the crosscutting nature of the associated concern, or just part of its broader, more complex implementation?
• Is the technique able to find similar concerns and is it able to easily distinguish between concerns that are not similar?

To answer these questions we believe that the focus of a mining technique should not be (only) on general symptoms but (also) on expected results (and specific symptoms): generic crosscutting concerns (sorts) that the technique expects to identify. This would allow for a clear assessment of the results (whether they are a relevant element of an expected sort instance) and the technique.

2.1. Sorts of crosscutting concerns

A crosscutting concern sort is a generic, atomic crosscutting concern described by a number of properties common to all its (concrete) instances like: (1) a general intent, (2) an implementation idiom in a non-aspect-oriented language (i.e., a specific symptom), (3) and a (desired) aspect mechanism to support the modularization of the sort’s concrete instances.

Table 1 shows several sorts from a longer list of proposed canonical sorts [4]. These are described by the three elements defining them as well as several examples of concrete instances.

Complex examples of crosscuttingness described in literature, like design patterns[2], are compositions of sort instances: an Observer pattern implementation, for example, involves instances of the role superimposition sort for the two defined roles (Subject and Observer), as well as instances of consistent behavior for the consistent actions of listeners notification and registration.

2.2. Defining “Target” crosscutting concern sorts for fan-in analysis

Fan-in analysis generates candidates based on the fan-in metric of a method: if a method is called from many, scattered places, the method is considered a potential seed. Hence fan-in is essentially a metric for the scattering symptom of the crosscutting concerns.

To focus our analysis, we define the targeted crosscutting concerns of fan-in analysis as instances of the Consistent behavior sort. The sort is shown in Table 1; it is described by an action and a number of (method) elements that consistently execute the action as part of their complete functionality. These elements, which are crosscut by the invocation of the specific action, are part of a context that can be formalized by a pointcut definition.

Concrete instances of the sort comprise credentials checks as part of the authorization mechanism, or logging of exception throwing events in a system.

2.3. (Candidate-)seed quality

We propose the (candidate-)seed quality as a measure for assessing the results of an aspect mining technique. The measure is defined as the percentage of elements in the (candidate-)seed that belong to the implementation of the crosscutting concern associated with the candidate, if any. The percentage of these elements gives a measure of the effort involved in reasoning and deciding about a candidate-seed.

In order to measure the (candidate-)seed quality, a technique has to describe the way one should reason about a candidate (i.e., the output of the technique); that is, we need to know what are the elements describing the candidate-seed and what is their relation with the targeted concerns.

Returning to the fan-in analysis example, the results of the analysis are described by the method with a high fan-in value and the callers of this method. However, not all the calls to the method with a high fan-in value are necessarily crosscutting, or part of a (same) Consistent behavior instance. A fan-in candidate for Consistent behavior is labeled as seed if its callers are part of a relationship (i.e., context definition) that can be formalized in a pointcut expression. The call reported by the technique crosscuts the elements in this context.

3. Candidate-seed quality for fan-in analysis

We discuss the quality measure for the fan-in technique and investigate what attributes could be relevant for improving it. The discussion uses for exemplification a number of selected results from the JHotDraw case-study.

JHotDraw 1 is an editor for 2D graphics developed as an open-source project and a show-case for how to apply design patterns [1]. We used the application as a case-study for fan-in analysis in a previous work, in which we also describe the concerns associated with the results of the analysis [3].

3.1. Seed quality for a number of concerns in JHotDraw

The selected examples from JHotDraw (typically) involve several crosscutting elements (i.e., sorts instances), like, for instance, Role superimposition and Consistent behavior instances for the Observer pattern implementation. We look at how the elements pertaining to Consistent behavior instances in these complex examples are identifiable by fan-in analysis, and what is the quality of the seeds for these instances.

1www.jhotdraw.org
The selection of the concerns is aimed at showing various relations between the elements (callers) provided by the technique for reasoning about its results.

### 3.1.1. Undo

The Undo concern involves around 30 classes like commands, tools, and (figure) handles elements. The changes spawned by the execution of the activities associated with these elements can be undone by specialized, dedicated UndoActivities. The UndoActivity classes are nested within their associated activities and implement the Undoable interface.

Fan-in analysis reports three seeds for the Undo concern, all of them part of instances of the targeted sort(s). The seeds are shown in Table 2. The first seed corresponds to a method implementing undo functionality and which is called by 24 methods. Most of the callers (22) are implementations of the undo method in the nested (UndoActivity) classes. The 22 callers follow the same idiom to invoke the reported seed: the invocation is the first call in the caller to check if the specific activity can be undone. The other two callers (UndoCommand.execute and UndoRedoActivity.redo) do not follow this idiom and the call is not part of a consistent behavior like for the other callers.

In order to decide about this candidate, we have to be able to observe specific symptoms of consistent behavior, like the structural relation between the first 22 callers to define a context, or the similar positions of the calls. The other two callers, on the other hand, make the analysis of the candidate harder, because they have to be investigated although they turn out not to be part of the consistent behavior concern. The relevant elements in the analysis of the callers are the first 22 callers and hence the quality is 22/24 (92%).

Similar, the quality for the other two candidates is 22/25 (88%) and 20/24 (83%), respectively.

### 3.1.2. Figure Change Observer

The Figure elements in JHOTDRAW participate in an implementation of the Observer pattern (Figure Change) by implementing the Subject role. A number of elements listen for changes in Figures by implementing the FigureChangeListener interface. The concrete figures implement or inherit the methods specific to the Subject role for allowing (de-)registration of listeners, and notification of changes in their state.

The (de-)registration action is a consistent behavior for the listeners of figure changes. This is implemented as a call to the method that adds/removes listener objects for a Figure: 

\[ Figure.add-/remove-FigureChangeListener \]

The notification of changes occurred in the Figures’ state is also a consistent behavior applying to elements changing this state. The concern is implemented as a call to the notification method: Figure.willChange, before making the changes, and Figure.changed, after the changes were made.

The consistent behavior for these instances of the sort is due to the relation between the callers and the callee in the context of the pattern: the listeners interested in changes have
to register in order to be notified, and the actions changing the state of the Subject object have to notify about this change. Hence all the callers of the reported seeds are relevant because the calls occur due to the participation into the pattern implementation. This participation is a secondary concern for the callers. Moreover, the concern implemented by the callees (the reported high fan-in methods) implies calls only from participants in the pattern, in the context of the pattern. The quality of these seeds is thus 100%.

Fan-in analysis reports several implementations of the described candidates. Table 2 shows the results omitting the multiple implementations that occur due to polymorphism.

The last reported method, `Figure.listener`, provides access to the listener reference in the class implementing the Subject role. The method is part of this role, however, it does not belong to a consistent behavior instance in the analyzed implementation of the Observer pattern.

### 3.2. Improving the quality of the fan-in candidates

To improve the quality of the fan-in candidates, we propose a number of properties to be considered for analyzing the callers of a candidate. These properties show possible relations between the callees of a method with a high fan-in, and are aimed at reducing the percentage of irrelevant elements describing a candidate, and hence reduce the effort of reasoning about the candidate.

The list of proposed properties comprises:

- structural relations between the callers. These relations include:
  - same hierarchy: The methods (callers) are defined by the same interface (/super class). As a particular case, the callees could be implementations of the same method. The (callers of the) seed for the `Undo` concern previously discussed falls into this category.
  - common roles: A method is associated with all the roles implemented by its class, so that the methods can share common roles. A role is typically defined by an interface. Methods that belong to the same hierarchy will also share the role that defines the hierarchy: the callers of the discussed `Undo` seed are declared by the `Undoable` interface, which defines the main role of the classes implementing the concrete callers of the seed. Similarly, the callers of the registration method in the previous Observer example are associated with the `FigureChangeListener` role. The main role for these callers is, however, defined by other interfaces, like the `Figure` interface.
  - – same class: The callers belong to the same class, as for the case of a class level contract.
  - etc.

- consistent call position: The position of the call, relative to the caller’s body, is consistent for the callers of the reported method with a high fan-in value. Such a case has been shown for the seeds discussed in Section 3.1.
- naming-based relations: The callers have similar names. The naming-based and the structural relations can be expressed by an AspectJ-like pointcut definition, while the call position could be an indication of the advice type (before/after).
- relations based on the structure of the call: similar or, sometimes, identical call sites. For example, Exception wrapping concerns typically consist of catching a specific type of exception and re-throwing an exception of a different type [3]. Another example consists of calls that occur together, like the notification of changes in the previously discussed instance of the Observer pattern for `Figure` changes: typically, a method changing the state of a `Figure` object would call `willChange` before the modifications, and then the `changed` method after completing the modifications.
- “intentional” relations between callers, such as modifiers of Subject objects in the context of the Observer pattern. The relations between the callees are due to their participation in the pattern implementation.

### 3.3. Results after extending fan-in with a callers analysis

Table 3 shows the quality of the fan-in seeds after analyzing the callers based on some of the proposed properties.
The mined instances that are part of the Undo concern are representative for many (most) of the concerns identified using fan-in analysis. In most cases, the quality of the seeds is improved by analyzing the proposed properties.

Although the original quality of the seeds for the Observer implementation is 100%, the analysis of the callers based on the proposed properties is meaningful due to the complex relation between the callers that has to be observed. Because the relation is intentional, it is expected that the investigated, mainly structural, properties to be irrelevant, offering little or no insight into the intentional relation. However, the structural properties can provide insight into the specific consistent behavior instances: Most of the callers (10 out of 11) of the registration method belong to classes that implement the listener role and that register themselves as listeners of a Figure object. The common role property groups and relates the 10 methods by this common, relevant concern.

A similar grouping and insight into the callers’ relation can be obtained by considering the call position property for the callers of the notification method.

The tool support for fan-in analysis, FINT ², allows for improving the quality of the results of the analysis. In the present version (v.0.5b), the user can display and investigate (1) relations between the callers of a method and their declaring types (i.e., common roles) or implementing classes, (2) relations between the callers and the position of the call, as well as (3) relations between the callers and all their callees.

Figure 1 shows a part of the tool support in FINT for the analysis of the callers. The user can investigate the relations between the callers of a method by their declaring interfaces: callers declared by the same interface are shown in a same color. Another analysis allows to look at the relations between the callers and the position of the call to the method with the high fan-in value. This position can be an absolute value or relative to the caller’s body.

4. Discussion and conclusions

To be able to assess aspect mining techniques, we need objective measures. In this work, we proposed a quality measure aimed at assessing the relevance of the results of an aspect mining technique and the effort required to analyze these results. We showed how this measure applies to fan-in analysis, and identified a number of properties to improve the quality of the results of this mining technique.

The proposed analysis of the callers to reason about a fan-in candidate also shows the potential relations between the elements describing a candidate. Such relations are important for defining the context of the concern through a pointcut construct.

The situations described for the Undo concern, where the callers are related by structural relationships, are common to many of the concerns discovered by fan-in analysis. These include Consistent behavior and Contract enforcement instances in JHotDraw, as well as concerns in other analyzed case-studies, like PETSTORE and TOMCAT [3]. The intentional relationship, on the other hand, are typically harder to detect. The intention of a developer could be captured by annotations in the code, which to be analyzed by aspect mining techniques.

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References

[1] E. Gamma, R. Helm, R. Johnson, and J. Vlissides. Design Patterns: Elements of Reusable Object-Oriented Software. Addison-Wesley, Reading, MA, 1994.

²http://swerl.tudelft.nl/bin/view/AMR/FINT