Spring 3-27-2018

Poster: Towards safe refactoring for intelligent parallelization of Java 8 streams

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Poster: Towards Safe Refactoring for Intelligent Parallelization of Java 8 Streams

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ABSTRACT

The Java 8 Stream API sets forth a promising new programming model that incorporates functional-like, MapReduce-style features into a mainstream programming language. However, using streams correctly and efficiently may involve subtle considerations. In this poster, we present our ongoing work and preliminary results towards an automated refactoring approach that assists developers in writing optimal stream code. The approach, based on ordering and typestate analysis, determines when it is safe and advantageous to convert streams to parallel and optimize parallel streams.

CCS CONCEPTS

• Software and its engineering → Software evolution; Automatic programming; Maintaining software;

KEYWORDS

refactoring, parallelization, typestate analysis, Java 8, streams

1 INTRODUCTION

Writing parallel programs can be difficult. MapReduce [1], a popular programming paradigm for writing a certain class of parallel programs, abstracts away much of this complexity by facilitating multi-node processing using succinct functional-like constructs.

Recently, mainstream languages such as Java 8 have adopted functional-style, MapReduce-inspired constructs for parallel and sequential data structure processing. In the case of Java, this functionality is embodied by the Stream API, introduced in Java 8 [4].

MapReduce, however, traditionally operates in a highly-distributed environment with no concept of shared memory, while Java 8 Stream processing operates in a single node under multiple threads or cores in a shared memory space. Since streams enable developers to pass behavioral parameters (λ-expressions) to collections for deferred execution, they can be easily executed either sequential or in parallel, making them especially attractive to those not normally familiar with functional programming. But, a burden is now placed on developers to manually determine whether running stream code in parallel results in an efficient yet interference-free program. Using streams correctly and efficiently requires many subtle considerations. In fact, ~4K questions on streams have been posted on Stack Overflow (http://stackoverflow.com/questions/tagged/java-stream), of which ~5% remain unanswered.

In general, these kinds of errors can lead to programs that undermine concurrency, underperform, and are inefficient. Moreover, these problems may not be immediately evident to developers and may require complex inter-procedural analysis, a thorough understanding of the intricacies of a particular stream implementation, and knowledge of situational API replacements. Manual analysis and/or refactoring (semantics-preserving, source-to-source transformation) to achieve optimal results can be overwhelming and error- and omission-prone as necessary changes can often be widespread. In this poster, we present our ongoing work and preliminary results in developing a novel automated approach, based on ordering and typestate analysis, that automatically identifies and executes refactoring opportunities where improvements can be made to Java 8 Stream code with the hope of this “intelligent” parallelization resulting more efficient, semantically-equivalent code. Although our preliminary data suggests that the approach is promising, speedup analysis is for future work.

2 MOTIVATION AND INSIGHT

Fig. 1 uses the Java 8 Stream API to process Widget collections. Fig. 1a is the original version, while Fig. 1b is the improved (but semantically equivalent) refactored version. A Collection of Widgets is declared (line 1) that does not maintain element ordering as HashSet does not support it. Note that ordering is dependent on the run time type rather than the compile-time type.

A stream, a data source view representing an element sequence supporting MapReduce-style operations, of unordered Widgets is created on line 2. This stream’s operations execute sequentially. Streams may also be associated with an encounter order (element visitation), which can be dependent on the stream’s source. In this case, it will be unordered since HashSets are unordered.

On line 3, the stream is sorted() by the corresponding intermediate operation, the result of which is a (possibly) new stream with
the encounter order rearranged accordingly. Intermediate operations are deferred until a terminal operation is executed (line 4). The execution results in a List of Widgets sorted by weight.

It may be possible to increase performance by running this stream’s “pipeline” (operation sequence) in parallel. Fig. 1b, line 2 displays the corresponding refactoring. Note, however, that had the stream been ordered, running the pipeline in parallel may actually result in worse performance due to the multiple passes and/or data buffering required by an operation like sorted(). Because the stream is unordered, the reduction can be done much more efficiently by breaking the problem into sub-problems [4].

In contrast, line 5 instantiates an ArrayList, which maintains element ordering. A Set of distinct widget weights is created beginning on line 6. Unlike previously, this collection takes place in parallel due to the corresponding call. Note though that there is a possible performance degradation here as distinct() may require multiple passes, the computation takes place in parallel, and the stream is ordered. Keeping the parallel computation but unordering the stream may improve performance but it is required to know whether it is safe to do so, which can be error-prone if done manually.

Our insight is that it may be possible to determine if it is safe to unorder a stream by analyzing the type of the resulting reduction. In this case, it is a collection to a Set, of which subclasses that do not preserve ordering exist. If we could determine that the resulting Set is unordered, unordering the stream would be safe since collecting into such a Set would not preserve ordering. The type of the resulting Set returned here is determined by the passed Collector, in this case, Collections.toCollection(TreeSet::new). Unfortunately, since the TreeSet will preserve the encounter order, we must keep the stream ordered. To improve performance here, it is advantageous to run this pipeline, perhaps surprisingly, sequentially. This transformation takes place in Fig. 1b, line 6.

3 OPTIMIZATION APPROACH

We propose several new refactorings, which include Convert Sequential to Parallel and Optimize Parallel Stream. The first deals with determining if it is advantageous (performance-wise) and safe (e.g., no race conditions, semantics alterations) to transform a sequential stream to one whose pipeline runs in parallel. The second deals with a stream that is already set to execute in parallel. The question here is what steps (i.e., transformations) can be taken to improve its performance, whether it is unordering the stream or converting the stream to execute sequentially.

Our in-progress automated refactoring approach involves using typestate analysis [2,5] to determine stream attributes when a terminal operation is issued. A typestate analysis variant is being developed since operations like sorted() return (possibly) new streams derived from the receiver with their attributes altered. To determine collection attributes, e.g., element ordering, a combination of points-to analysis and reflection is used, with the former to interprocedurally approximate return value run time types, and the latter to instantiate the class to obtain ordering data. This is viable as collections do not normally alter ordering during object lifetime.

Our generalized typestate analysis works by tracking the state of stream instances using the two labeled transition systems (LTSs), one of which tracks execution mode and the other ordering. Stream typestate is then merged with that of “intermediate” streams to obtain the final typestate at the terminal operation since that is when all of the (queued) intermediate operations will execute.

4 PRELIMINARY RESULTS

Our refactoring approach has been implemented as an Eclipse plug-in and built upon WALA (http://wala.sf.net). A preliminary experiment on 11 open source Java projects demonstrates that our tool promisingly deems ∼33% of 128 total streams refactorable, i.e., those passing our preconditions. Determining whether the refactoring results in more optimal code is part of our future work. Major reasons that streams are not refactorable include λ-expressions side-effects (∼45%) and that the reduction ordering is preserved by the target collection (∼22%, cf. §2). Although speedup analysis is for future work, it has been shown that a similar manual refactoring can improve performance [3, Ch. 6].

5 CONCLUSION & FUTURE WORK

We have outlined our work-in-progress towards an automated refactoring approach that “intelligently” optimizes Java 8 stream code. The approach, based on ordering and typestate analysis, automatically deems when it is safe and advantageous to run stream code either sequentially or in parallel. In the future, we will expand our corpus and formulate a transformation algorithm.

REFERENCES